

## Supplemental Materials

# AeroTrans: Hourly AOD Retrieval over Land from MSG-1/SEVIRI Imagery Integrating Transformer and Transfer Learning

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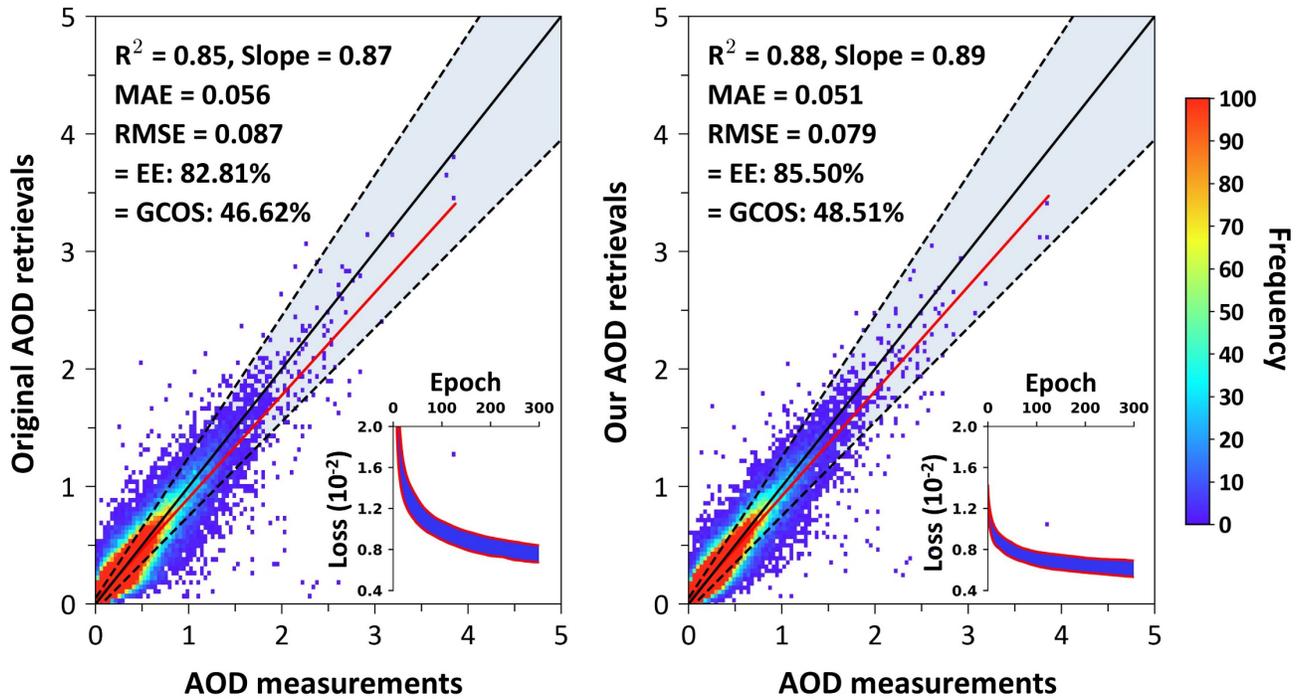
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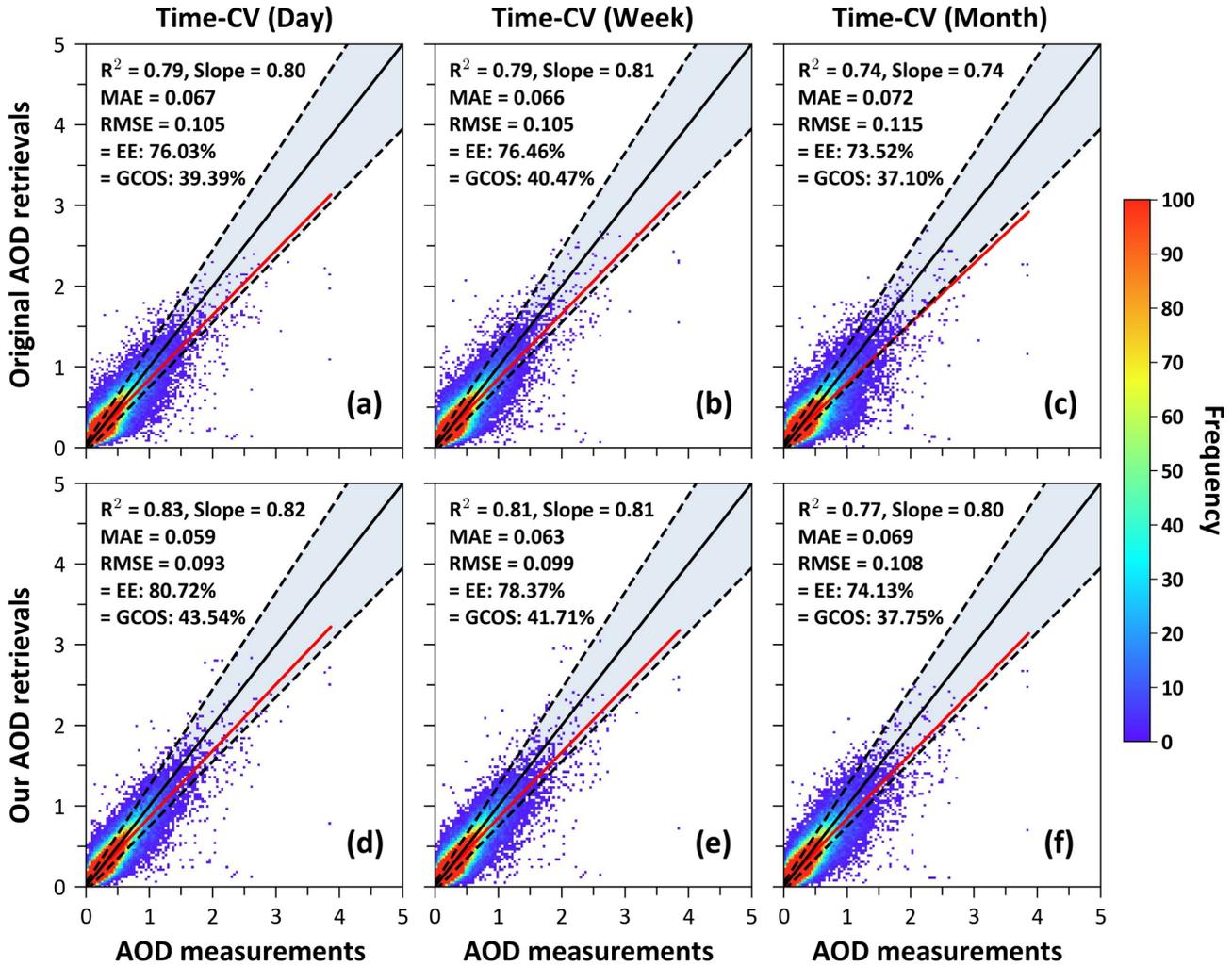
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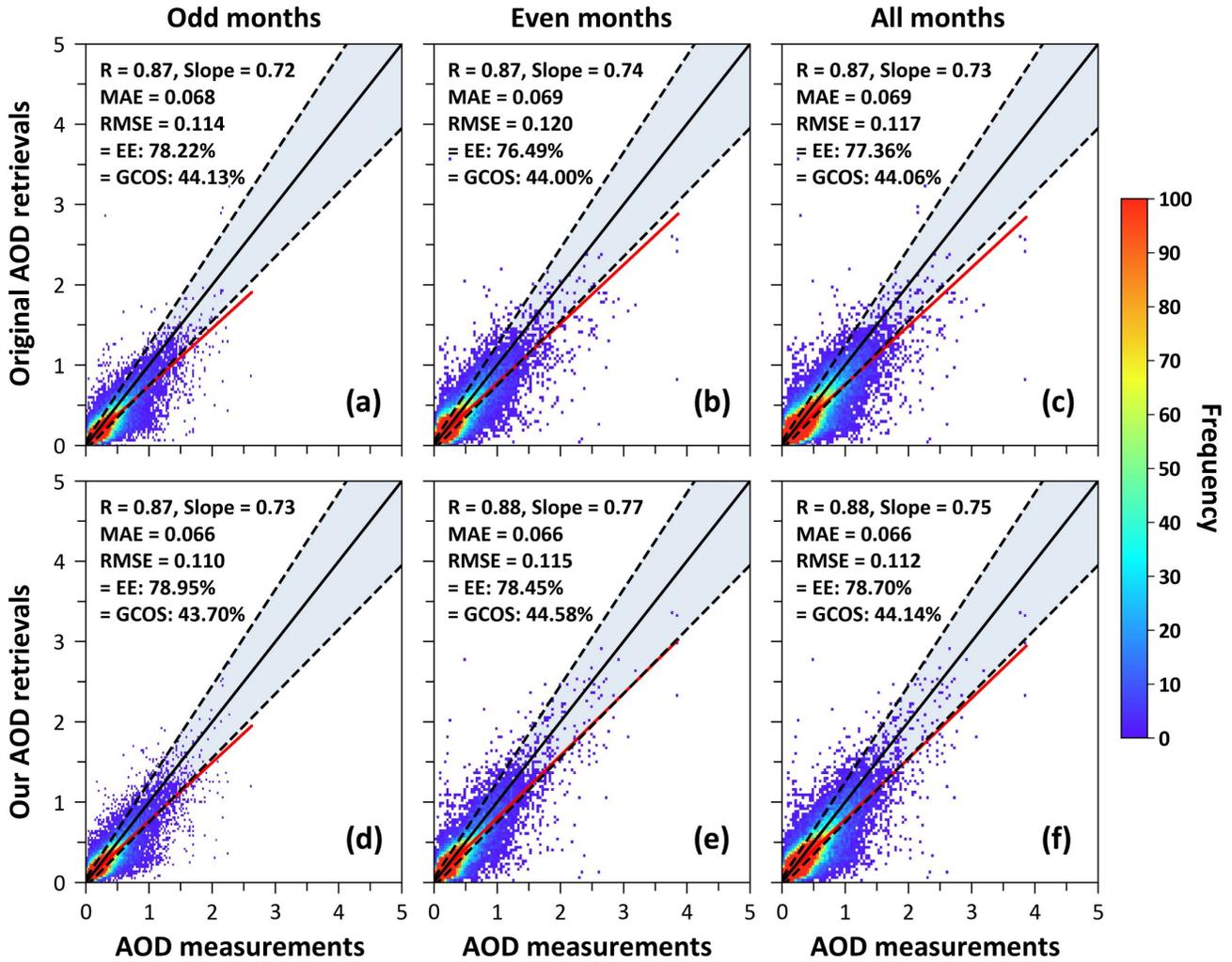
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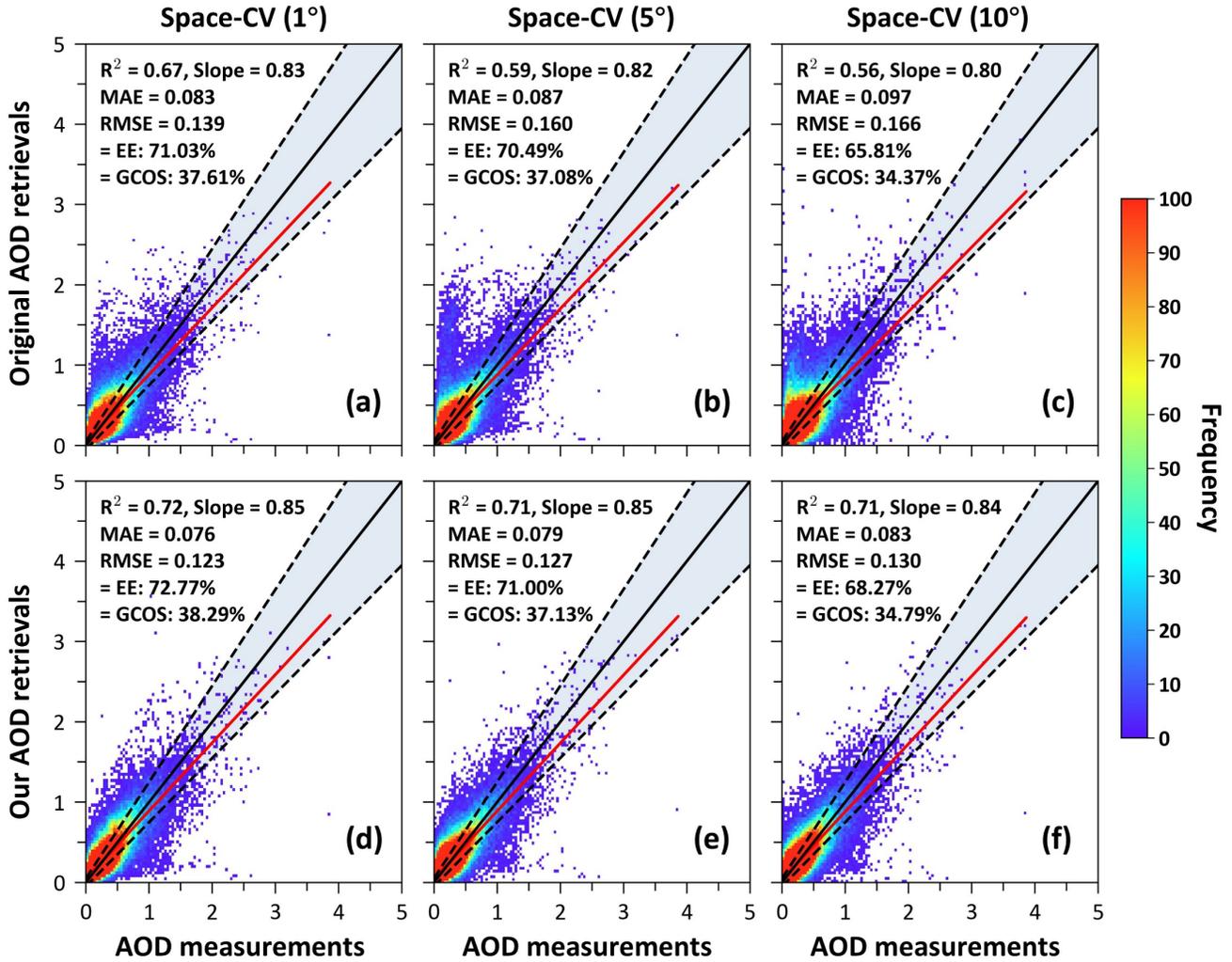
**Figure S1.** Density scatter plots of (a) original model (trained directly without pre-training) and (b) our model (pre-trained and fine-tuned) for hourly AOD retrievals against ground-based measurements, assessed using sample-based ten-fold cross-validation approaches. In each panel, the black dotted line represents the expected error range, the black solid line indicates the 1:1 line, the red solid line shows the linear regression line, and the subgraph illustrates the fluctuation of loss values ( $10^{-2}$ ) over the training process. The numbers in the top-left corner of the subfigure represent the coefficient of determination ( $R^2$ ), slope of the regression line, mean absolute error (MAE), root mean square error (RMSE), percentage of retrievals within the expected error (EE), and Global Climate Observing System (GCOS).



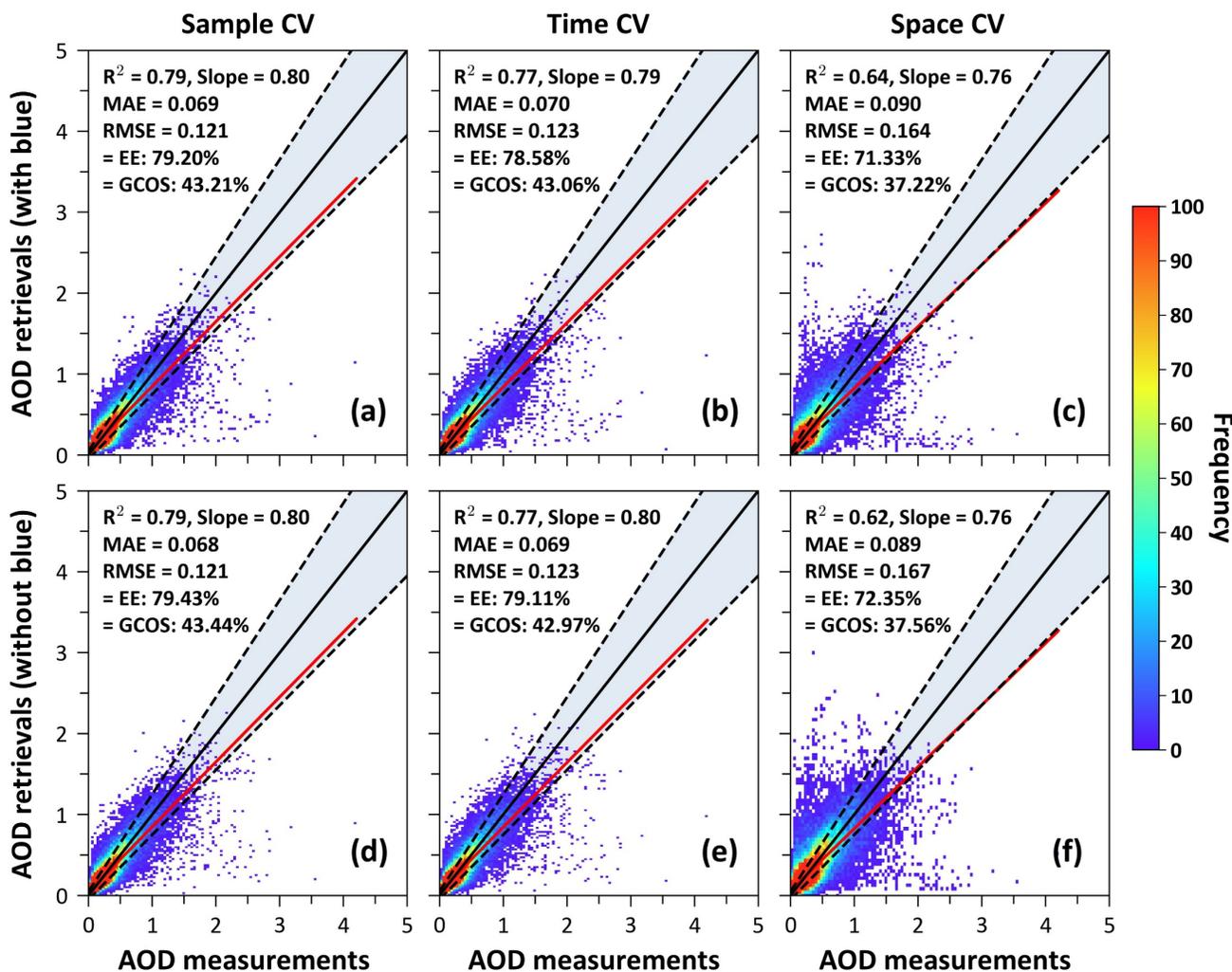
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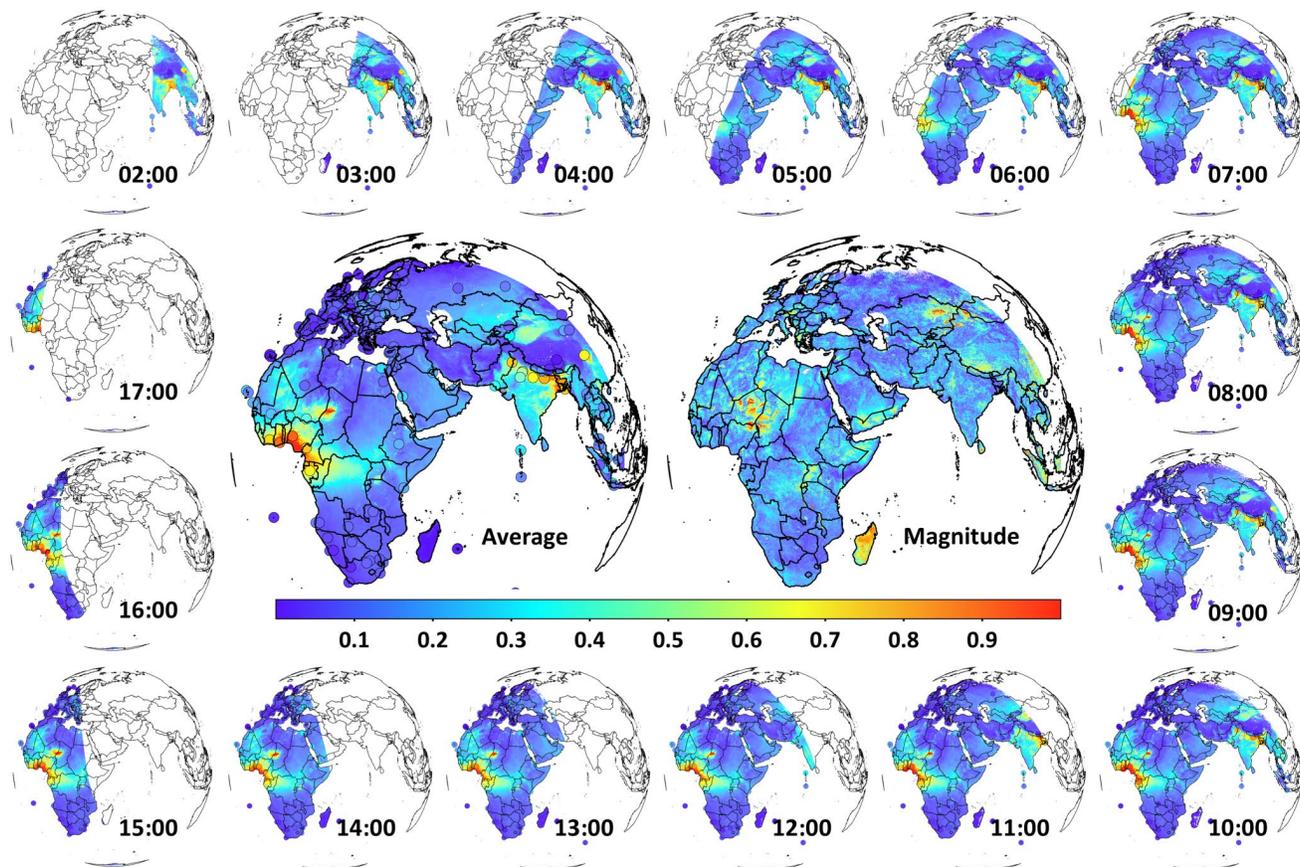
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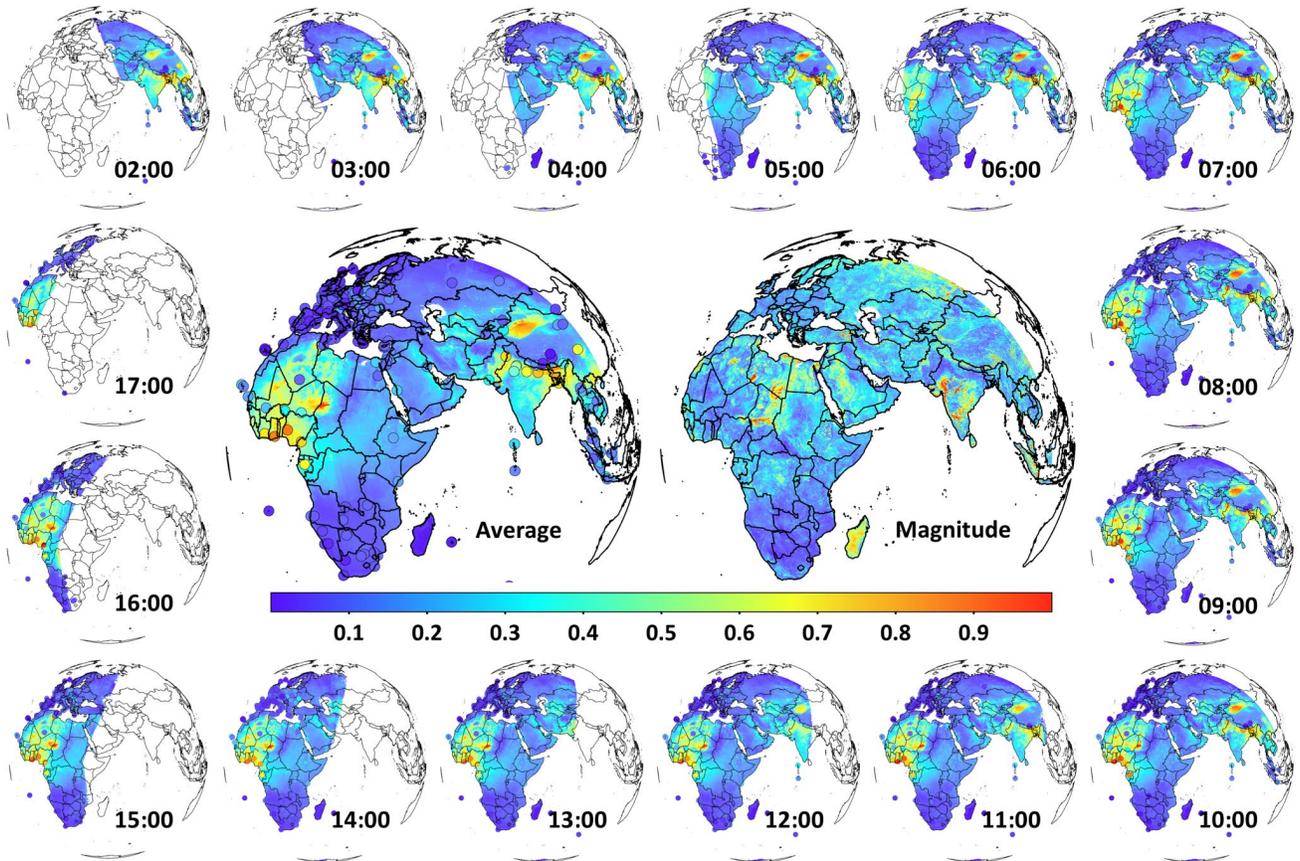
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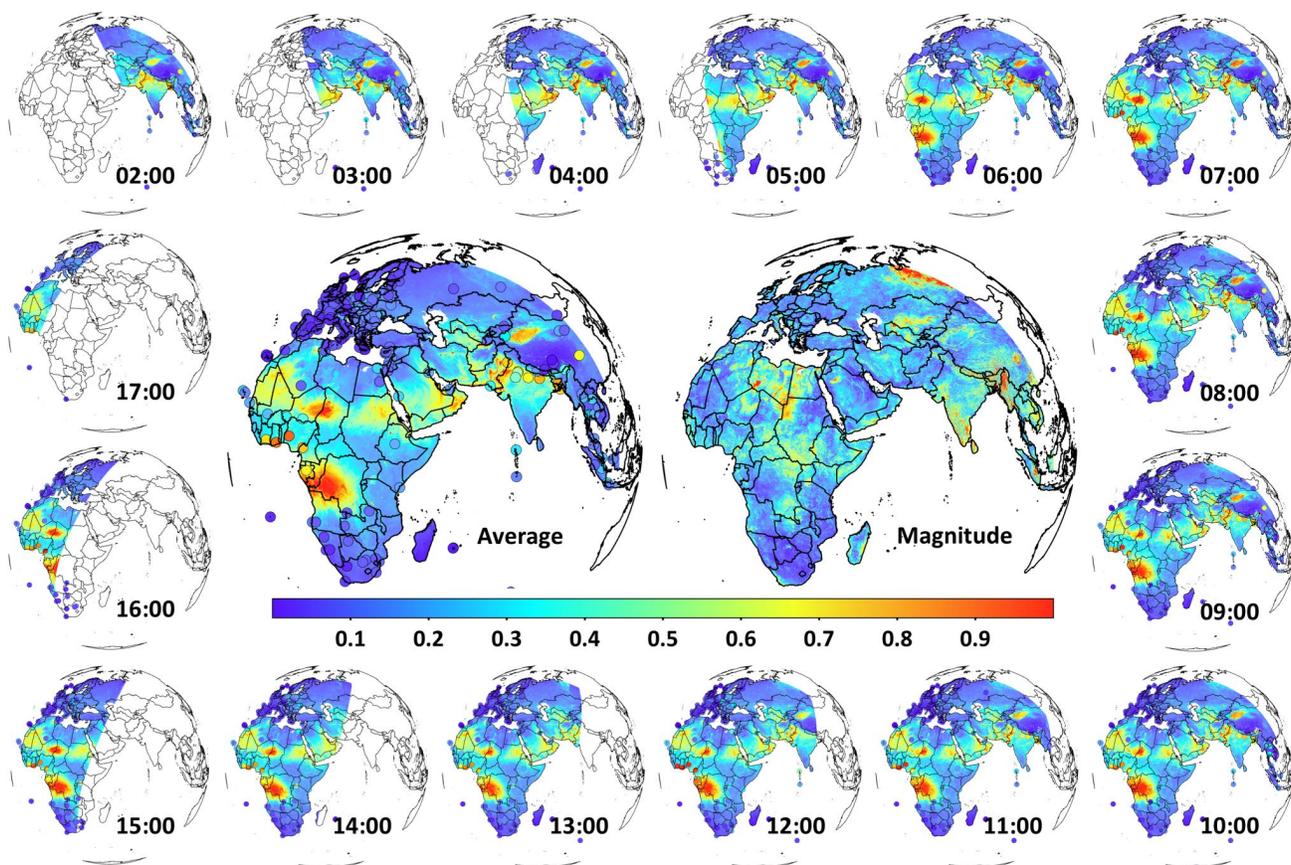
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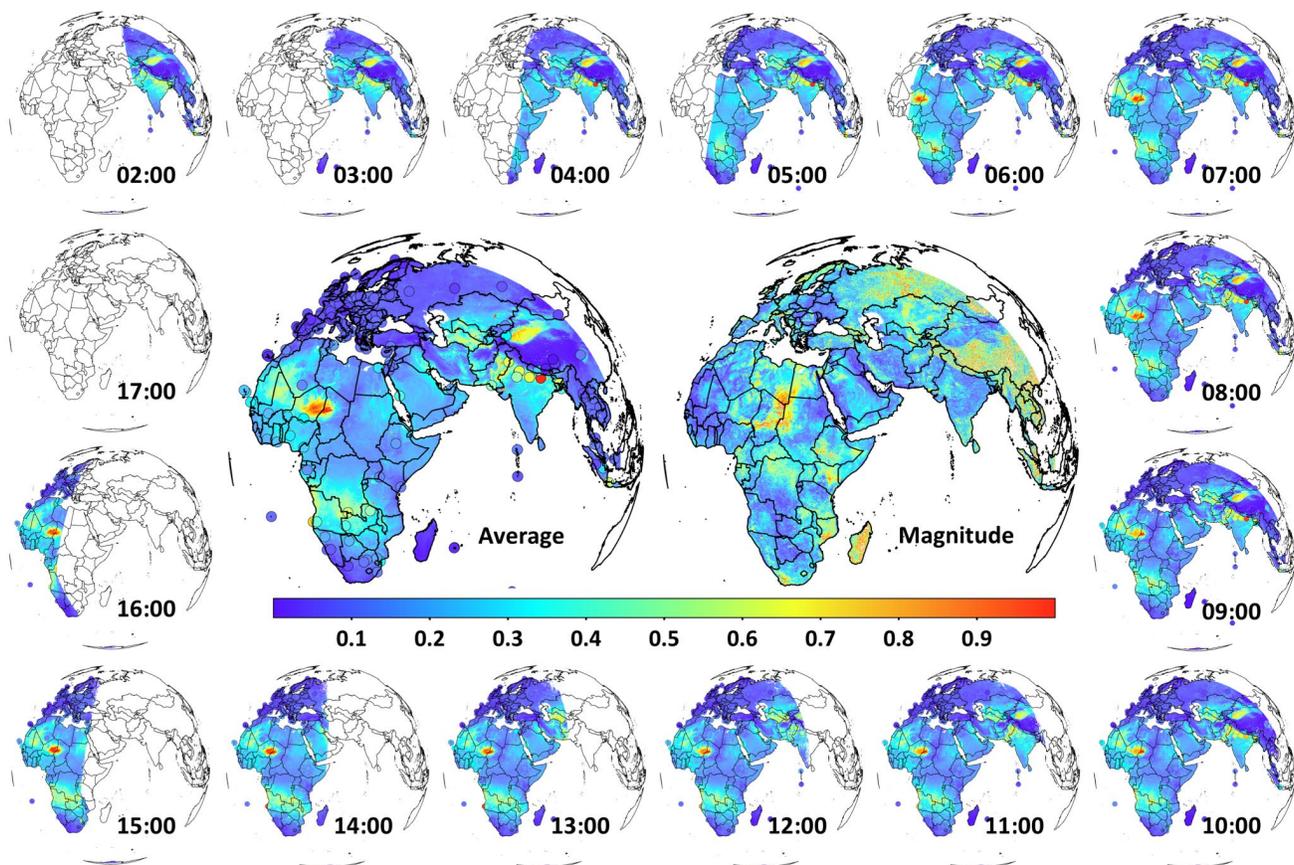
**Figure S6.** Diurnal variations in AOD at 3-km spatial resolution from 02:00 to 17:00 UTC over Europe, Africa, and West Asia, together with maps of the daily mean AOD and the amplitude of diurnal variation for the Northern Hemisphere winter (December–February).



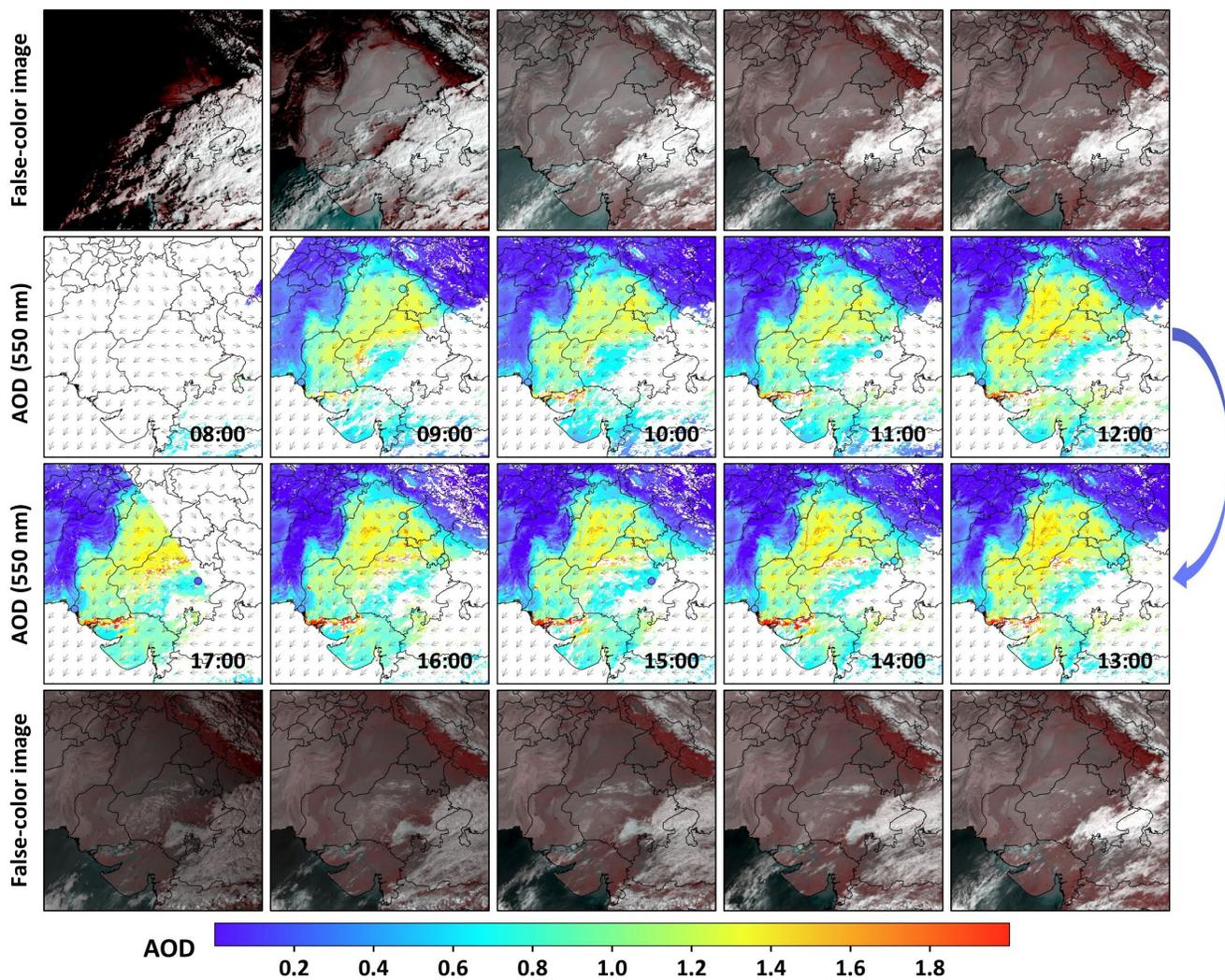
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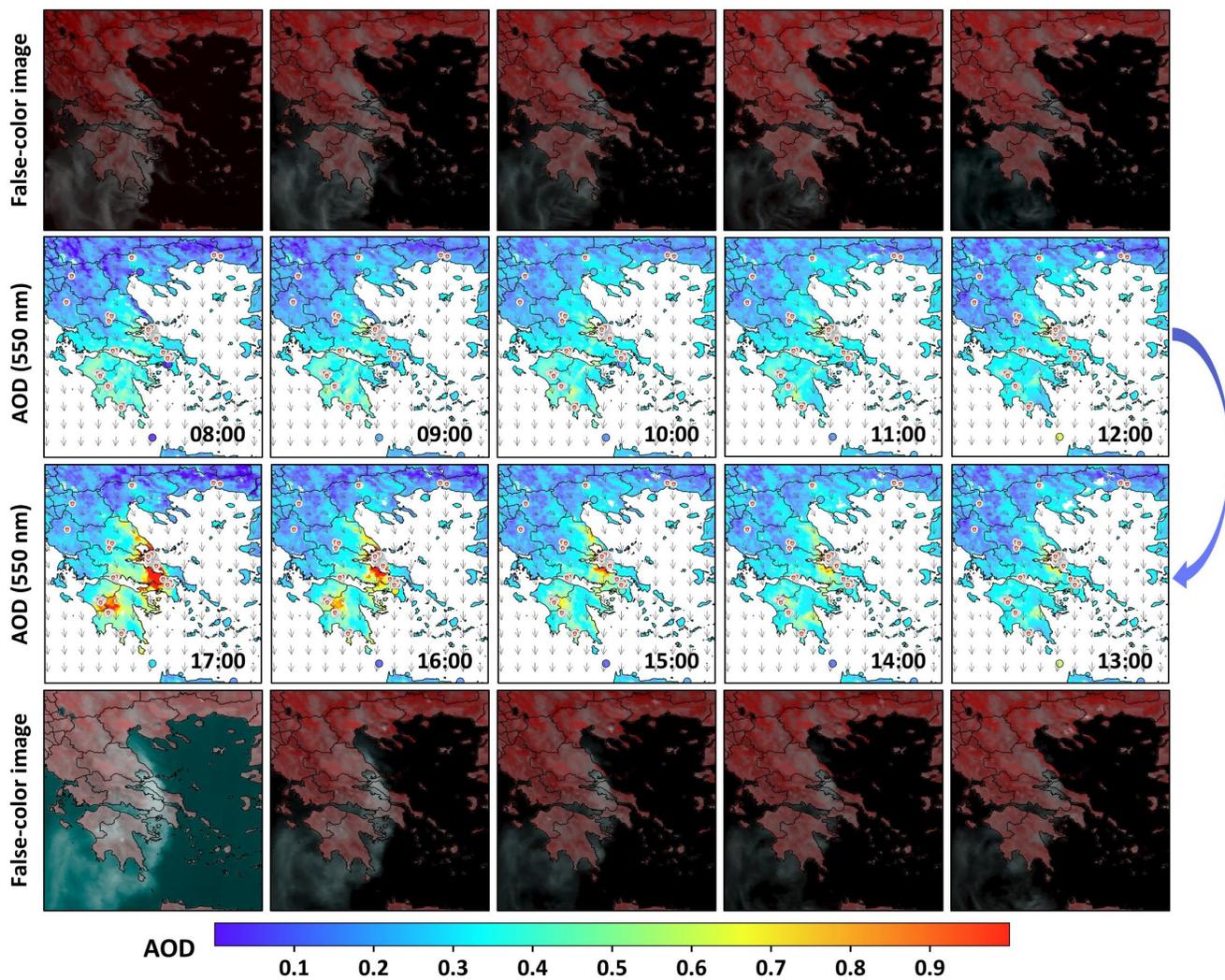
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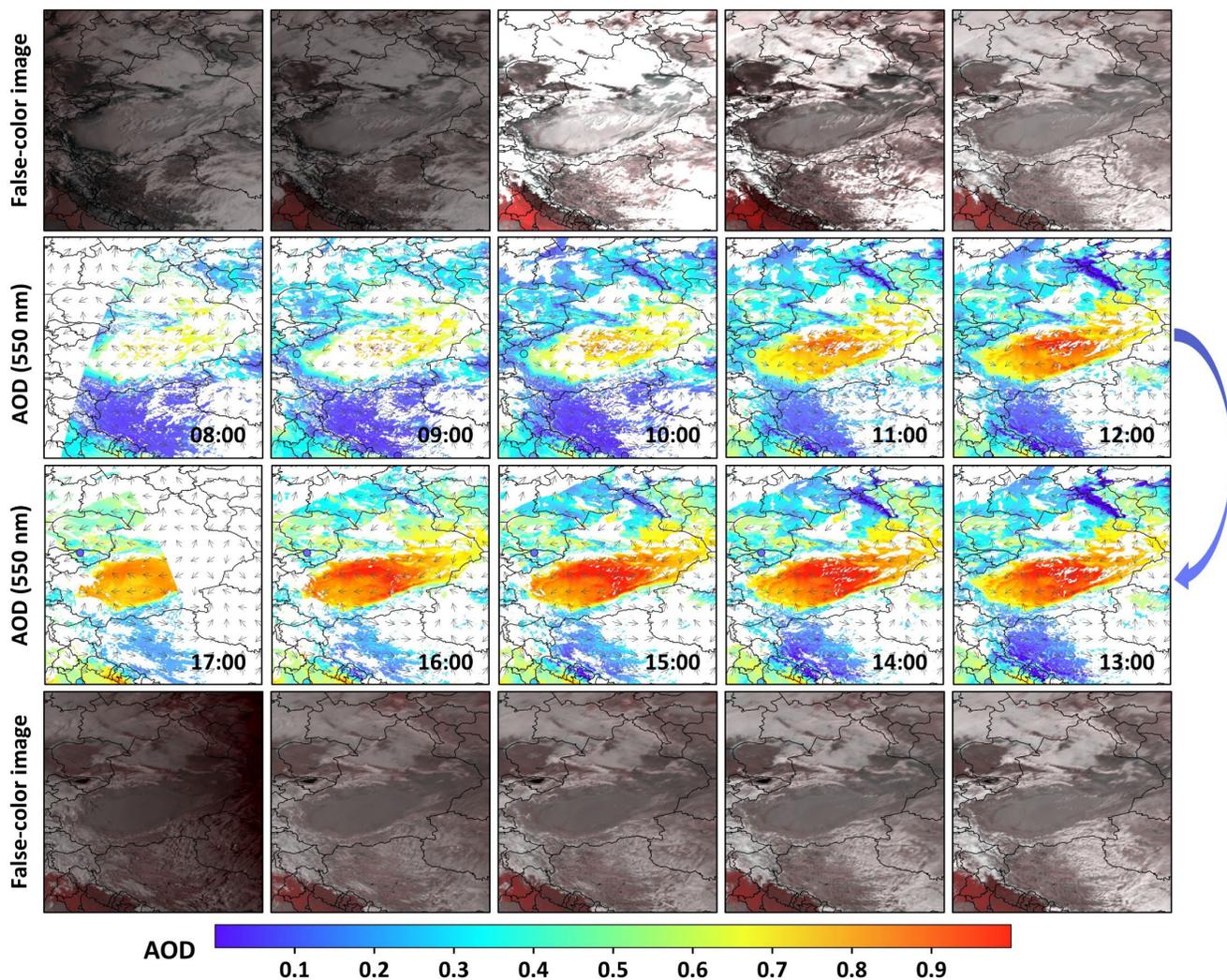
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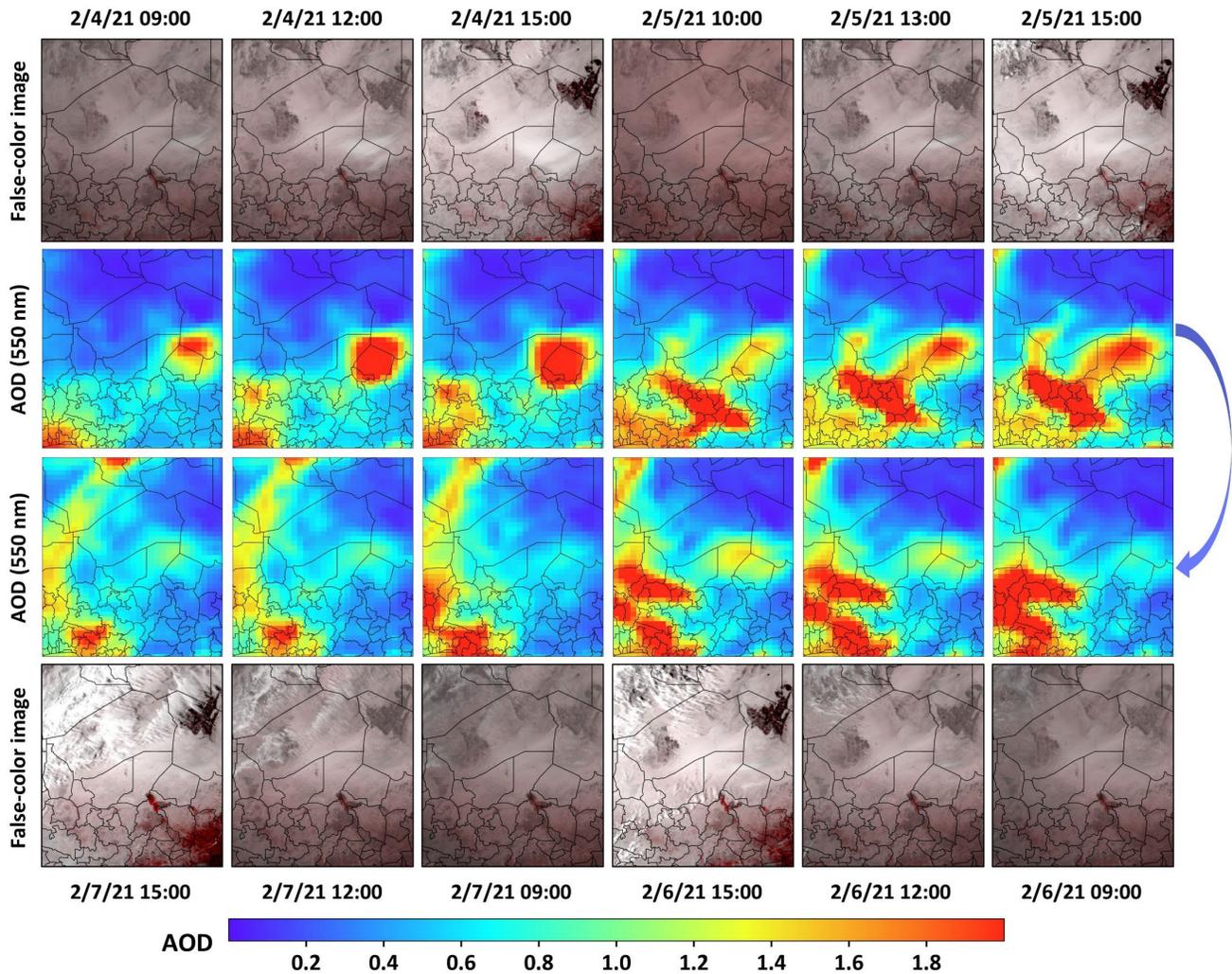
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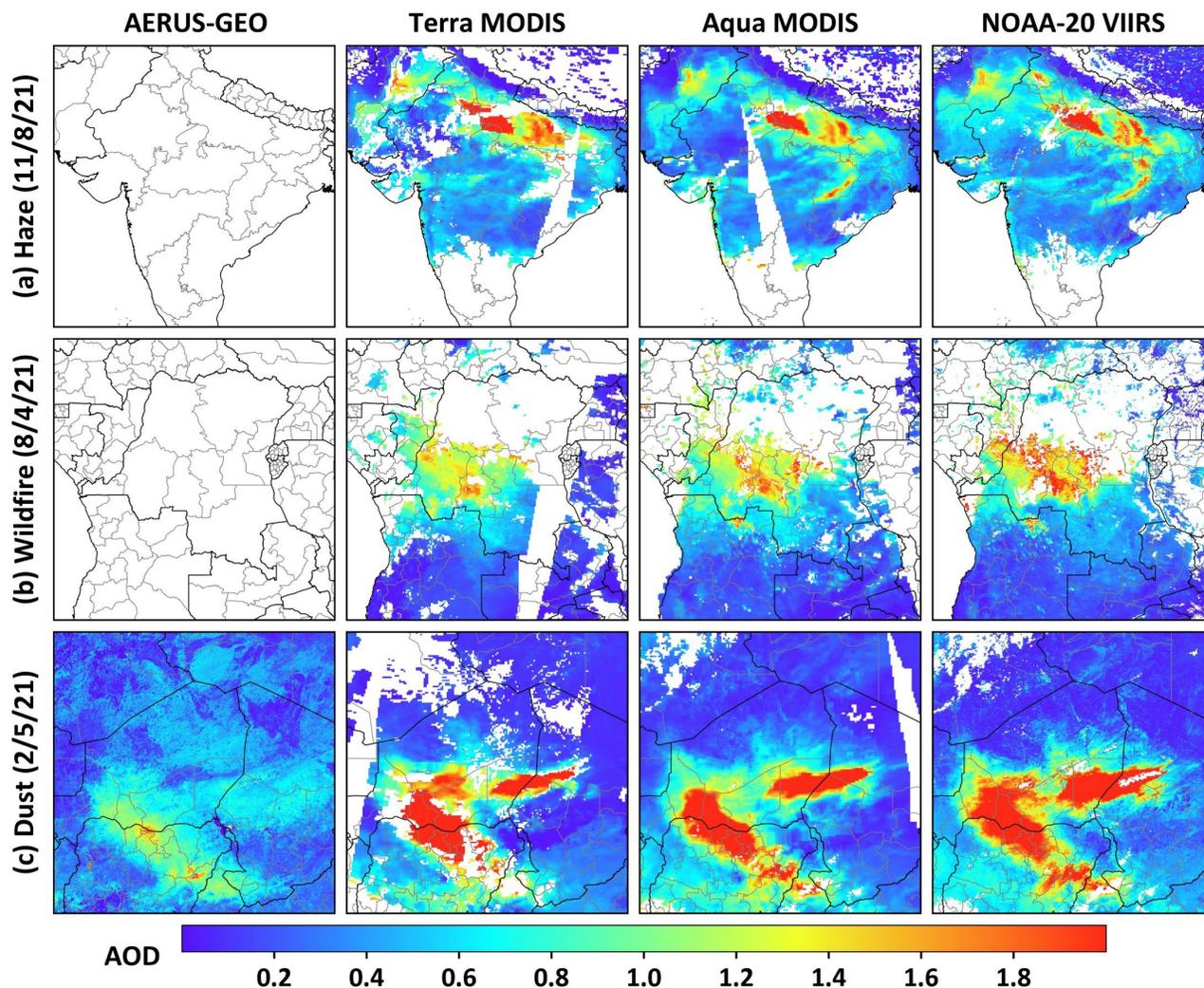
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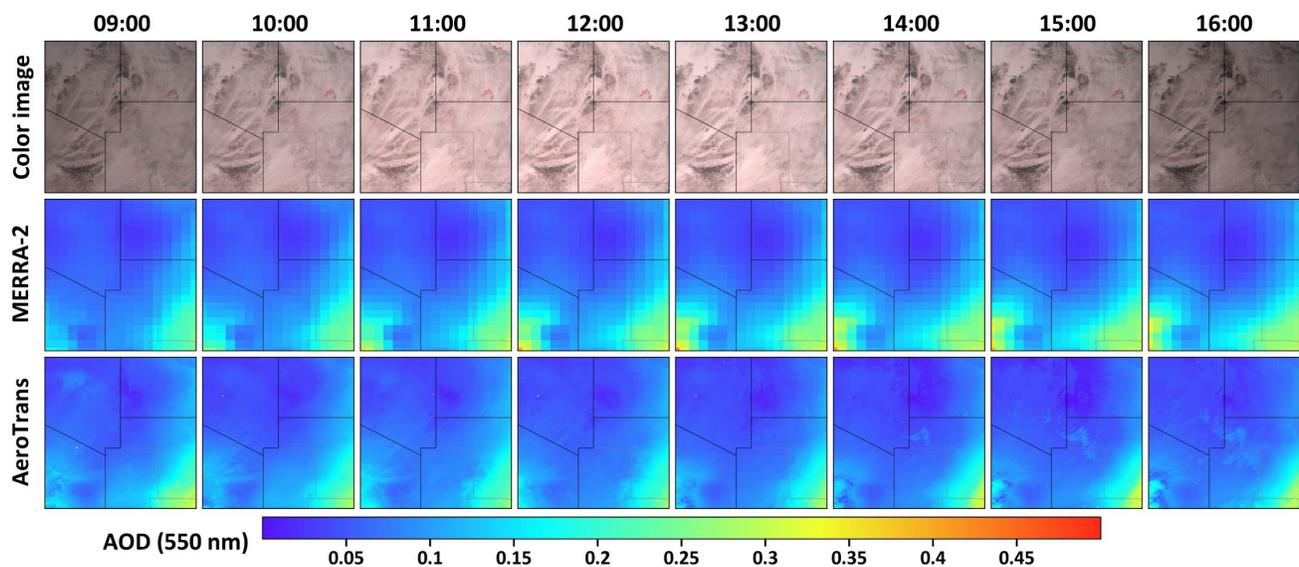
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**Figure S14.** Spatial distributions of available AOD retrievals from MSG-4/SEVIRI AERUS-GEO (635 nm, 3 km, daily-average), Terra/MODIS DB (550 nm, 10 km, ~10:30 local time, LT), Aqua/MODIS DB (550 nm, 10 km, ~13:30 LT), NOAA-20/VIIRS DB (550 nm, 6 km, ~13:30 LT) products during (a) a haze event on November 8, 2021 over India ( $67^{\circ}\text{E}$ – $89^{\circ}\text{E}$ ,  $11^{\circ}\text{N}$ – $33^{\circ}\text{N}$ ), (b) a wildfire event on August 4, 2021 in central and western Africa ( $10^{\circ}\text{E}$ – $35^{\circ}\text{E}$ ,  $17^{\circ}\text{S}$ – $8^{\circ}\text{N}$ ), and (c) a dust storm event on February 5, 2021 in the Sahara Desert ( $2^{\circ}\text{E}$ – $22^{\circ}\text{E}$ ,  $7^{\circ}\text{N}$ – $24^{\circ}\text{N}$ ). The absence of MSG-4 results in (a) and (b) is due to a lack of retrievals during the study period.



**Figure S15.** Spatial distributions of hourly AOD over northwestern Africa ( $20^{\circ}\text{E}$ – $30^{\circ}\text{E}$ ,  $16^{\circ}\text{N}$ – $26^{\circ}\text{N}$ ) on February 26, 2021, derived from the MERRA-2 reanalysis and our AeroTrans-MSG model, in a region without available ground-based measurements.

**Table S1.** Summary of spectral band parameters for the MSG/SEVIRI multispectral instrument.

Band name	Spectral range ( $\mu\text{m}$ )	Center wavelength ( $\mu\text{m}$ )	Spatial resolution (km)
HRV	0.40–1.10	0.75	1
VIS0.6*	0.56–0.71	0.64	3
VIS0.8*	0.74–0.88	0.81	3
NIR1.6*	1.50–1.78	1.64	3
IR3.9*	3.48–4.36	3.92	3
WV6.2	5.35–7.15	6.25	3
WV7.3	6.85–7.85	7.35	3
IR8.7	8.30–9.10	8.70	3
IR9.7	9.38–9.94	9.66	3
IR10.8	9.80–11.80	10.80	3
IR12.0	11.00–13.00	12.00	3
IR13.4	12.40–14.40	13.40	3

\* Indicates bands adopted in this study for AOD retrieval using deep learning

**Table S2.** Hyperparameter settings for our AeroTrans-MSG model.

Classification	Hyperparameters	Values
Model architecture*	Model input/output	22/1
	num_layers	2
	d_model	22
	nhead	1
	layer_norm_eps	1e-5
	feed-forward dimension	128
	dropout	0.01
	activation	ReLU
	device	CUDA
	Training parameter	optimizer
loss function		MSE
learning-rate schedule		halved after 100 epochs; batch size
batch size		1024
total training epochs		500
early stopping		patience = 20 with MSE threshold = 0.001

\*Note: The Transformer Encoder model is implemented using PyTorch's torch.nn package with other hyperparameters that use the framework defaults unless specified.