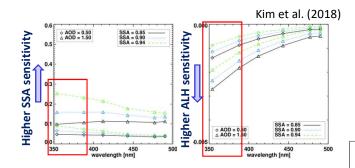


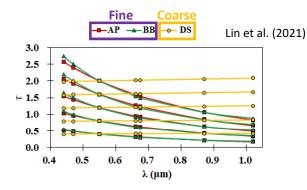
Minseok Kim, Jhoon Kim, Hyunkwang Lim, Seoyoung Lee, and Yeseul Cho Dept. of Atmospheric Sciences, Yonsei University, Republic of Korea

Remote sensing synergy of multiple sensors

Sensor spectral range affects sensitivity & spatial coverage of aerosol retrieval

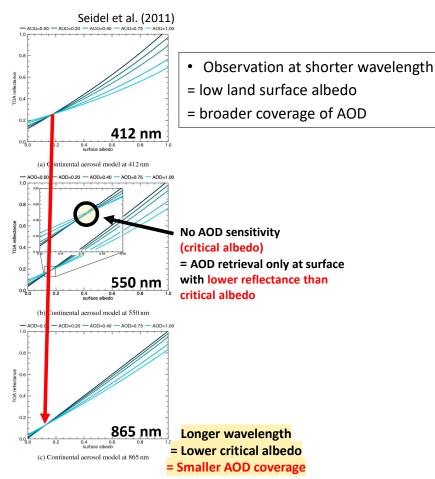


! UV: sensitive to aerosol absorption & height



! VIS-NIR: sensitive to aerosol size

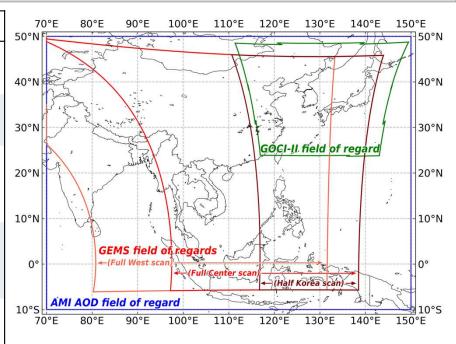
- High sensitivity = retrieval of the optical property.
- Low sensitivity = robust retrieval against the optical property.
 - different uncertainty characteristics



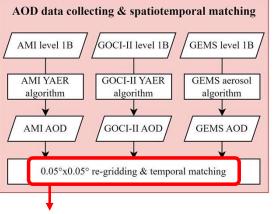
Introduction Data Methodology Results Summary

GK-2A, -2B / GEMS, AMI, GOCI-II aerosol algorithm

	АМІ	GEMS	GOCI-II
Satellite	GeoKompsat-2A (GK-2A)	GeoKompsat-2B (GK-2B)	
Purpose	Meteorology monitoring	Environment monitoring	Ocean color monitoring
Spectral range	VIS~ <mark>IR</mark> (0.47-13.3 μm) 16 <mark>band</mark>	UV~VIS (300-500 nm) 0.6 nm res. hyperspectral	VIS-NIR (0.38-0.86 μm) 13 <mark>bands</mark>
Field of regard	<mark>Full-disk</mark> , East Asia, Korean local area	East Asia (5°S~45°N, 75°E~145°E)	Korean local area
Spatial resolution	0.5 km (<mark>Red</mark>), 1 km (Green, Blue), 2 km (IR)	3.5 km × 8 km	<mark>250 m</mark>
Temporal resolution	<mark>10 min.</mark> (Full-disk), 2 min. (EA, KLA)	1 hour (8 times during daytime)	1 hour (10 times during daytime)
Aerosol product	Yonsei aerosol retrieval (YAER) algorithm 6 km res.	GEMS V2 operational algorithm 3.5 km × 8 km res.	Yonsei aerosol retrieval (YAER) algorithm 2.5 km res.



L2 AOD fusion strategy

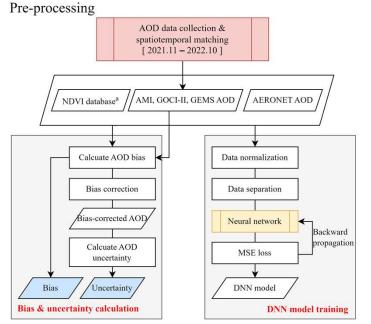


Spatial matching
 GEO coordinate → 0.05° ×0.05° grid
 Inverse distance weighting of maximum 3 neighboring pixels of each grid points.

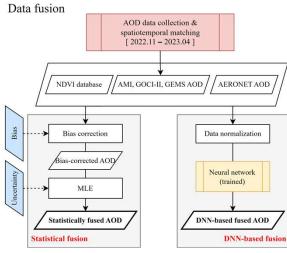
Neural network Input Fully connected Batch normalization ReLU Fully connected Batch normalization ReLU Fully connected Batch normalization ReLU Fully connected Batch normalization

Output

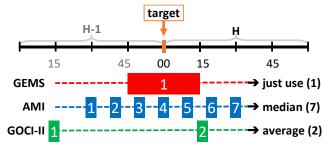
<Uncertainty estimation with 1-year data> : used AERONET V3 LEV2.0



<Data fusion for 6 months> : validated with AERONET V3 LEV1.5

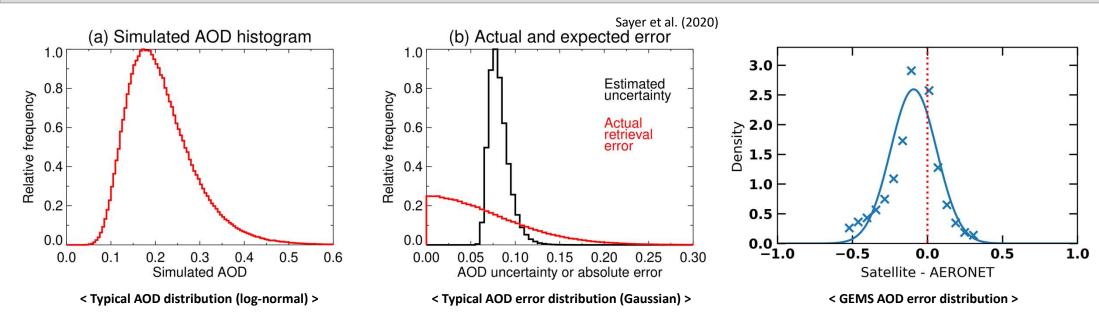


Temporal matching



Introduction Data Methodology Results Summary

Statistical fusion: 10 bias correction



Typical AOD has Gaussian error distribution (mean of zero, standard deviation of unity)

- ⇒ Gaussian fitting with actual retrieval data returns **non-zero** mean, and **non-unity** standard deviation.
- ⇒ non-zero mean = bias / non-unity standard deviation = uncertainty
- ⇒ A likelihood function of the MLE fusion is based on the assumption of unbiased satellite retrieval.
- ⇒ Retrieval bias needs correction.

Statistical fusion: 2 Maximum Likelihood Estimation

Maximum Likelihood Estimation

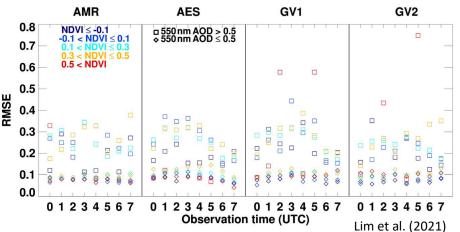
A likelihood function that assumes a **Gaussian error** having $\mu = 0$, $\sigma = uncertainty(R)$

$$\rho(x) = \sum_{i} \frac{1}{R\sqrt{2\pi}} \left(-0.5 \frac{\tau_{MLE} - \tau_i}{R_i}\right),\,$$

- ho(x) : likelihood function
- τ_{MLE} : target(fused) AOD
 τ_i : AOD of instrument i
- R_i : uncertainty (RMSE)

 au_{MLE} maximizing the likelihood function

$$\tau_{MLE} = \frac{\sum \tau_i R_i^{-2}}{\sum R_i^{-2}},$$



Why consider "pixel-level" uncertainty?

Satellite AOD retrieval have different uncertainty characteristics based on various error sources such as (a)surface condition, (b)light path, (c)aerosol loading, (d)aerosol type.

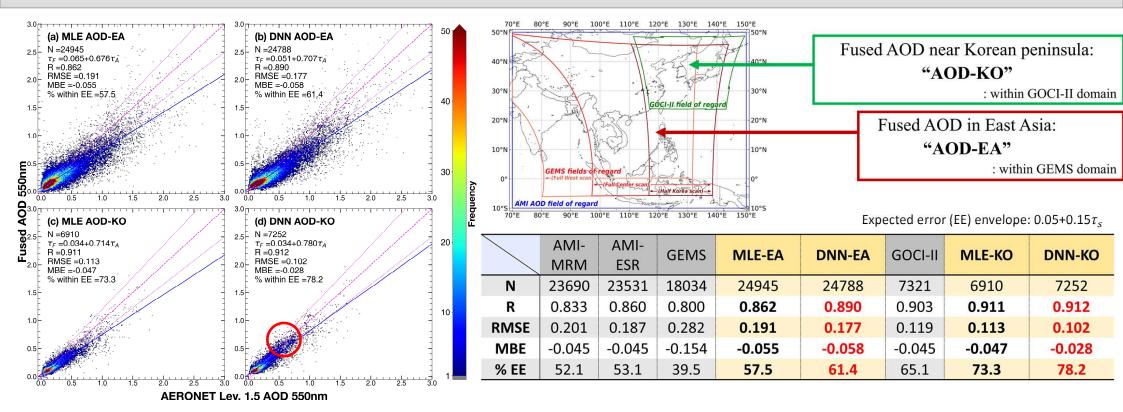
Uncertainty that reflects retrieval condition of each pixel is needed.

■ How?

Calculate uncertainty(RMSE) at different (a)NDVI, (b)observation time, (c)AOD, (d)aerosol type

Introduction Data Methodology Results Summary

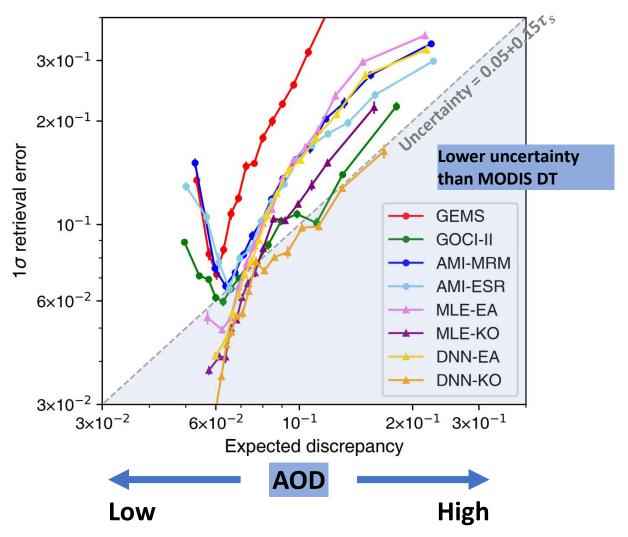
Validation with ground-based (AERONET) AOD



- Underestimation of high aerosol loading affects MLE AOD.
 - Better fusion result is expected after version update of GEMS aerosol algorithm.
 - Using neural network mitigates the problem.

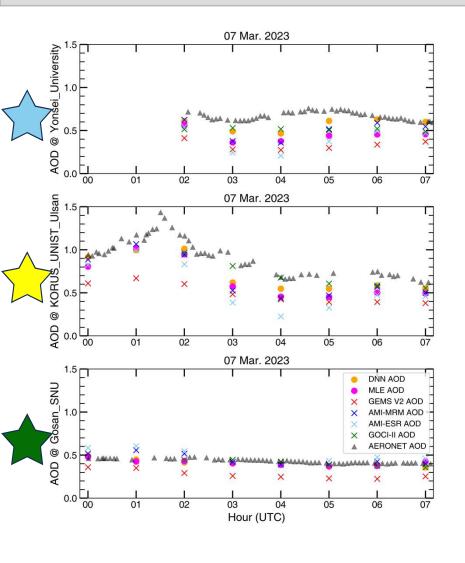
- Bias correction enhances low AOD accuracy of MLE AOD.
 - High MLE-KO AOD's % within EE.
 - % within EE of DNN-KO AOD is much higher due to better quality at higher AOD

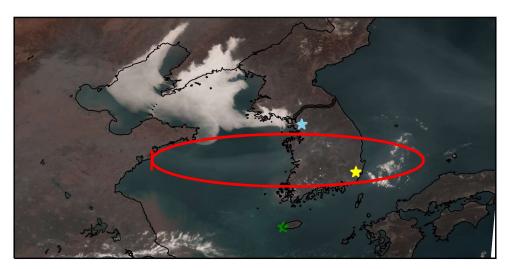
Performance Evaluation with MODIS DT expected error



- "Expected discrepancy": **AOD uncertainty** estimated based on MODIS DT expected error (EE: $0.05+0.15\tau_s$)
- Expected discrepancy is linearly related to AOD.
- Compare AOD uncertainty with one standard (MODIS DT).
- Within –EA region, MLE and DNN uncertainties are similar.
 - Merging GEMS + AMI-MRM + AMI-ESR, the uncertainty result at high AOD is similar to AMI-MRM.
 - Uncertainty of fused AOD products are low at low AOD.
- DNN-KO AOD has the lowest uncertainty.

Performance Evaluation with MODIS DT expected error





- (Yonsei_University) all AOD products were underestimating.
 - AOD value of DNN AOD is closer to AERONET, but diurnal variation is better captured with MLE AOD
- (KORUS_UNIST_Ulsan) AOD > 1.0 is observed with AERONET.
 - Both MLE and DNN AOD well captured high AOD.
- (GOSAN_SNU) Diurnal error variation of AMI-MRM, AMI-ESR AOD
 - Wrong diurnal AOD variation of AMI is mitigated after fusion.

Summary

- Bias and RMSE of GEMS, AMI, GOCI-II AOD products are calculated from Nov. 2021 to Oct. 2022.
- Three AOD products are fused from Nov. 2022 to Apr. 2023.
- Compared to EE based on MODIS DT, fused AOD performs better at low aerosol loading.
- At higher AOD, DNN algorithm works better than statistical fusion.
- Underestimation of AOD at higher aerosol loading affects fused AOD product.
- Aerosol data fusion stabilizes diurnal error variation.

Reference

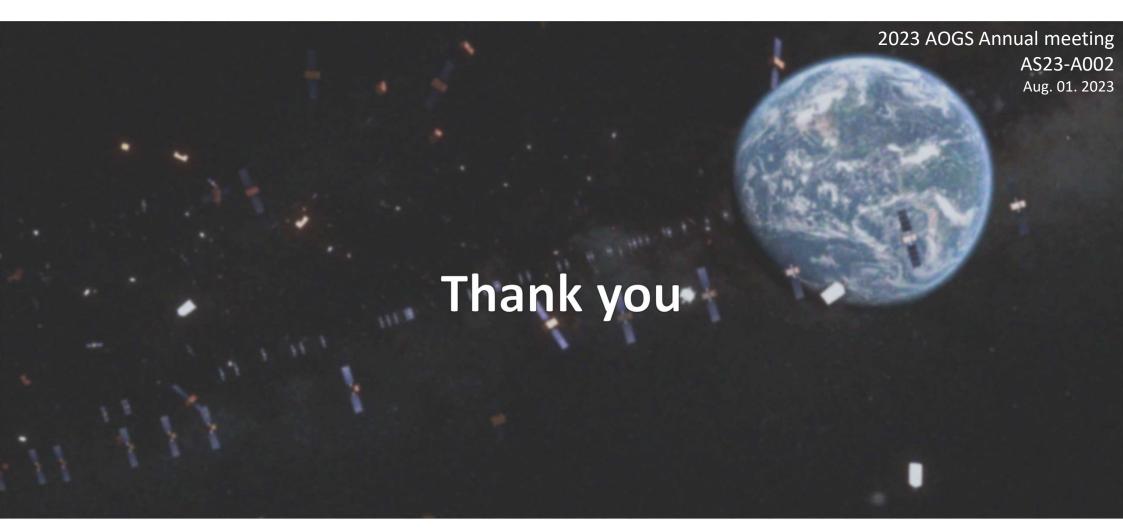
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