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# AOD fusion based on pixel-level uncertainty using geostationary satellite instruments: GEMS, AMI, and GOCI-II

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# Introduction Data Methodology Results Summary GK-2A, -2B / GEMS, AMI, GOCI-II aerosol algorithm

	ΑΜΙ	GEMS	GOCI-II	70°E 80°E 90°E 100°E 110°E 120°E 130°E 140°E 150°E 50°N
Satellite	GeoKompsat-2A (GK-2A)	GeoKompsat-2B (GK-2B)		40°N
Purpose	Meteorology monitoring	Environment monitoring	Ocean color monitoring	30°N 30°N 20°N 10°N 0° 6EMS field of regards 0° 6EMS field of regards 0° 10°S 70°E 80°E 90°E 100°E 110°E 120°E 130°E 140°E 150°E
Spectral range	VIS~ <mark>IR</mark> (0.47-13.3 μm) 16 <mark>band</mark>	UV <mark>~</mark> VIS (300-500 nm) 0.6 nm res. <mark>hyperspectral</mark>	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.	2







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.

 $\Rightarrow$  Retrieval bias needs correction.

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Summary

Introduction

## Data Methodology

## **Statistical fusion: ② Maximum Likelihood Estimation**

#### Maximum Likelihood Estimation

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

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

- $\rho(x)$  : likelihood function  $\tau_{MLE}$  : target(fused) AOD
- $\tau_{MLE}$  : target(fused) AOD •  $\tau_i$  : AOD of instrument *i*
- **R**<sub>i</sub> : uncertainty (RMSE)

 $au_{\textit{MLE}}$  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



Results



- 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



Results

#### Summary

### 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.

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