

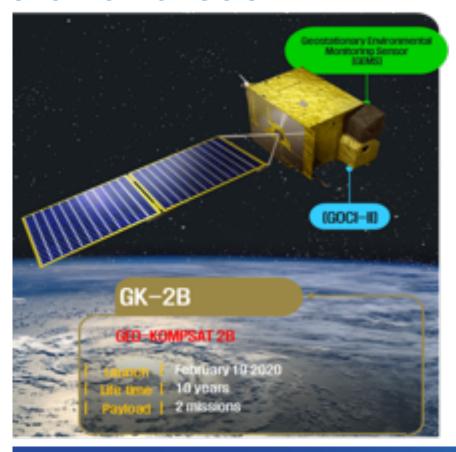
GOCI-II, current status, validation and applications

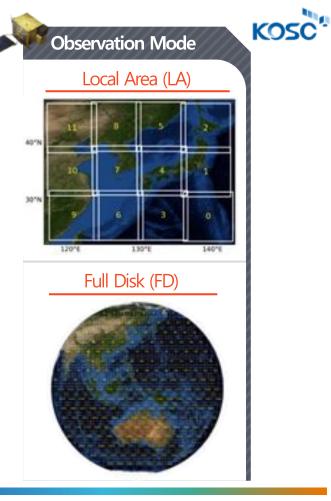
Korea Ocean Satellite Center, KIOST /
National Ocean Satellite Center, KHOA
Jong-Kuk Choi and HeeYoon Park (with all KOSC and NOSC staffs)





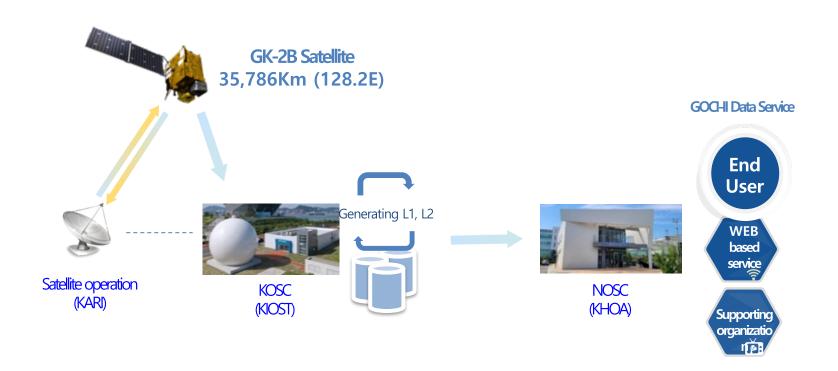
Overview of GOCI-II





Public Service for GOCI-II





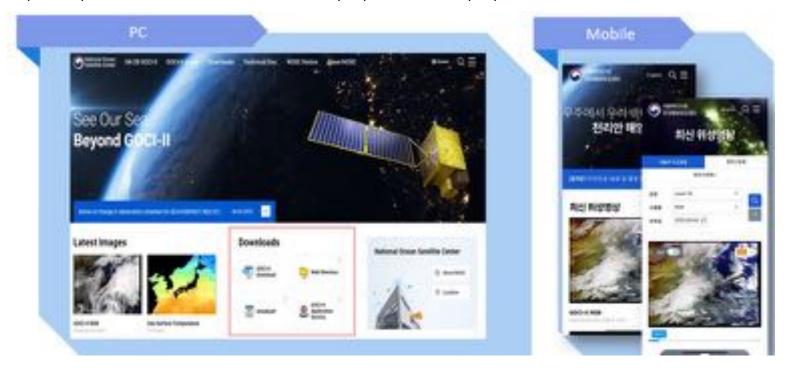
* National Ocean Satellite Center (NOSC) in Korea Hydrographic and Oceanographic Agency (KH

Public Service for GOCI-II by NOSC



Established new website (www.nosc.go.kr) and started public service from 2021

GK2B(GOCI-II) Web Downloads **7-fold increase** : ('21) 1.2million → ('24) 8.5million



Project on GOCI-II accuracy enhancement





Establishing Cal/Val standardization and improving accuracy at international level

Development of technology for cal/val of GOCI-II products

Research on algorithm improvement for GOCI-II products Development of atmospheric correction technique based on the integration of GeoKompsat-2A/2B

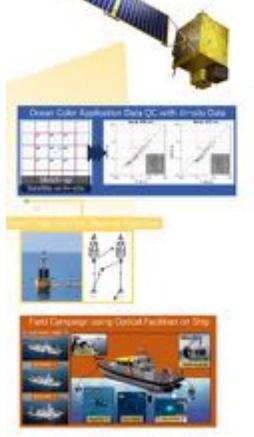


II. Works we are doing

Collection of CAL/VAL data









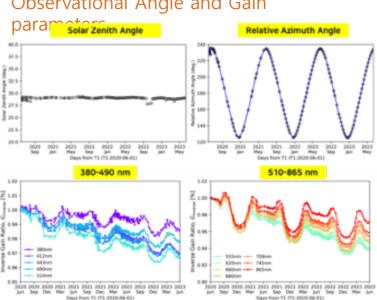
GOCI-II Radiative Calibration

$$G_{inverse}(B,t) = \frac{\bar{G}(B,t_0)}{\bar{G}(B,t)}$$



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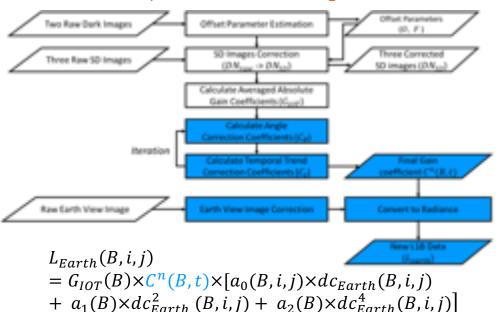
Time series of Solar Diffuser (SD) Observational Angle and Gain



- $G_{inverse}$ has seasonality
- $G_{inverse}$ has decreasing trend



Flow chart of Improved GOCI-II RC algorithm



 $G_{IOT}(B)$: absolute gain coefficient,

 $C^{n}(B,t)$: GOCI-II correction coefficient,

dc: corrected digital number after dark image correction

 a_0, a_1, a_2 : coefficient of GOCI-II RC algorithm

Improvement of GOCI-II RC algorithm





GOCI-II Radiometric Correction Algorithm

☐ Final GOCI-II RC equation

$$L_{TOA}(B, i, j)$$

$$= G_{IOT}(B) \times C_{\theta}^{n}(B, \theta) \times C_{t}^{n}(B, t) \times G_{inverse}(B, t)$$

$$\times [a_{0}(B, i, j)dc(B, i, j) + a_{1}dc^{2}(B, i, j)$$

$$+ a_2 dc^4(B, i, j)$$

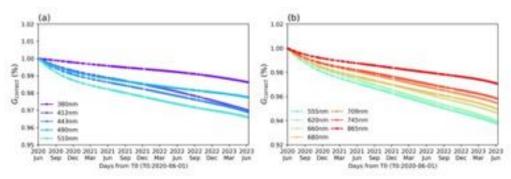
□ Angle correction model

$$C_{\theta}(B, \theta_{RAA}) = b_0 + \sum_{i}^{3} b_i(B) \times \sin(\theta_{RAA})^i$$

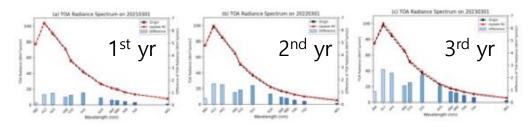
☐ Temporal trend correction model

$$C_t(B,t) = c_0 + \sum_{j=0}^{5} c_j(B) \times t^j$$

Timeseries of GOCI-II Corrected Gain Trend



Effect of RC: TOA radiance



Improvement of GOCI-II RC algorithm





GOCI-II Radiometric Correction Algorithm



Poster #: 39

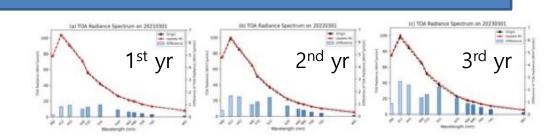
Date: Monday

Session #: 1

Title: Onboard Radiometric Calibration of GOCI-II Using Solar Diffuser and Long-Term Monitoring of Diffuser Stability

☐ Temporal trend correction model

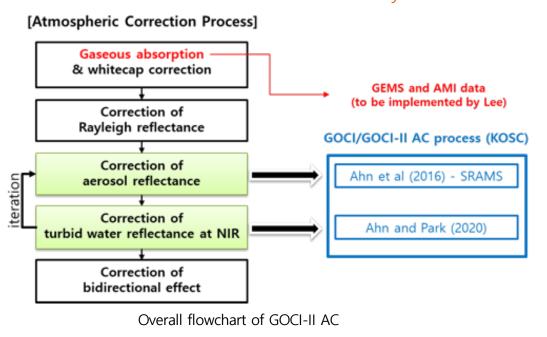
$$C_t(B,t) = c_0 + \sum_{j=0}^{5} c_j(B) \times t^j$$



Atmospheric correction and it's Cal/Val

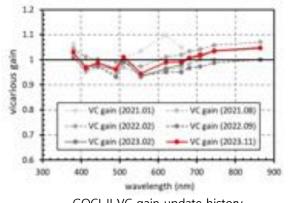


GOCI-II AC and VC methods are theoretically based on the SeaWiFS algorithm



[Vicarious Calibration Process]

- Current GOCI-II's vicarious calibration relies on the R_{rs} data derived from VIIRS in case-I waters
- Machine learning with simulation dataset for case-I water is used for spectral conversion of VIIRS R_{rs} into GOCI-II's R_{rs}



GK-2A/B Fusion for GOCI-II gas absorption correction



■■ Improvements of GOCI-II gas absorption correction through via fusion with GK-2A/B data

* G2GS Dataset: NCEP[H₂O, O₃] , OMI [NO₂]

** GK-2 Dataset : AMI [H₂O, O₃] , GEMS [NO₂]

GK-2B/GOCI-II



TCO: Total column ezone, TPW: Total precipitable water TNO₂: Total column nitrogen dioxide

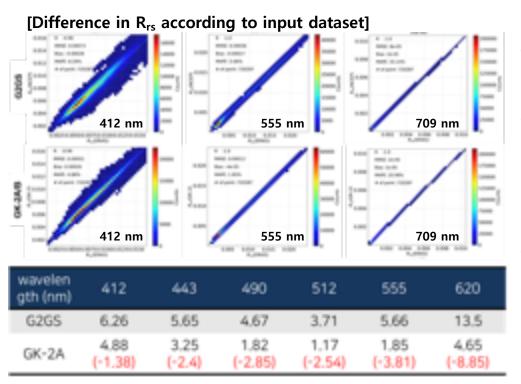
- AMI & GEMS: same geostationary orbit as GOCI-II
- valuable input sources of for gas absorption correction
- provide real-time trace gases data with higher spatia resolution (~8 km) and accuracy than NCEP data (0.25°)

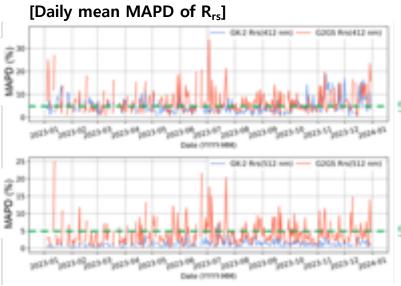
	Absorbing gas	Validation metrics	G2GS dataset* (used in G2GS)	GK-2 dataset**	Reference
		RMSD (mm)	10.2	3.65	
	Water vapor	MAPD (%)	35.13	13.01	RAOB data
	·	R	0.82	0.98	
		RMSD (DU)	18.01	9.54	
	Ozone	MAPD (%)	4.34	2.46	ERA-5 reanalysis
		R	0.82	0.95	
		RMSE (DU)	0.436	0.311	Dondonio
al	Nitrogen dioxide	MAPD (%)	71.3	56.56	Pandonia global network
		R	0.0	0.7	data

GK-2A/B Fusion for GOCI-II gas absorption correction



Comparison with an ERA5-based reference to assess the impact of input data on R_{rs}





 GK-2A/B Fusion: Improve accuracy and strengthens the temporal stability of error in R_{rs}

GK-2A/B Fusion for GOCI-II gas absorption correction



Comparison with an ERA5-based reference to assess the impact of input data on R_{rs}

Poster #: 46

Date: Monday

Session #: 1

Title: Improvements of GOCI-II gas absorption correction through via

fusion with GK-2A/B data

wavelen gth (nm)	412	443	490	512	555	620
G2GS	6.26	5.65	4.67	3.71	5.66	13.5
GK-2A	4.88 (-1.38)	3.25 (-2.4)	1.82 (-2.85)	1.17 (-2.54)	1.85 (-3.81)	4.65 (-8.85)

 GK-2A/B Fusion: Improve accuracy and strengthens the temporal stability of error in R_{rs}

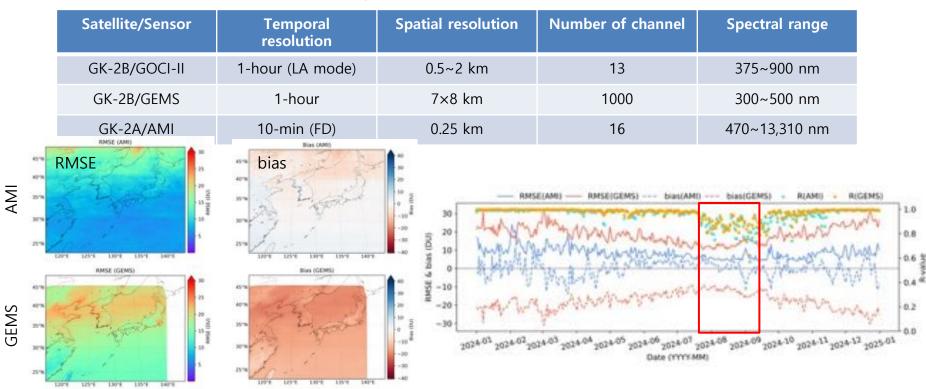
, 23, 44, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224, 46, 224

Comparison of AMI and GEMS TCO for GOCI-II Atmospheric Correction



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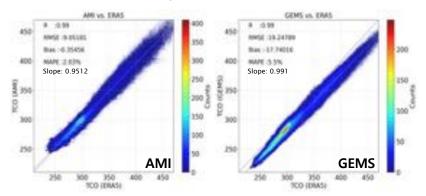
Sensor Characteristics and TCO Accuracy Assessment over the GOCI-II Local Area *TCO: Total Column Ozone



Comparison of AMI and GEMS TCO for GOCI-II Atmospheric Correction

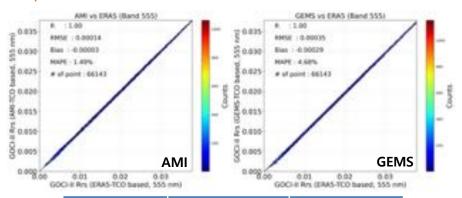


Evaluation of AMI/GEMS TCO



	AMI	GEMS
R	0.99	0.99
RMSE	9.05181	19.24789
Bias	-0.35456	-17.74016
MAPE	2.03 %	5.5 %

Impact of AMI/GEMS TCO on **GOCI-II Rrs** Retrieval



	АМІ	GEMS
R	1.00	1.00
RMSE	0.00014	0.00035
Bias	-0.00003	-0.00029
MAPE	1.49 %	4.68 %

→ The AMI data is considered to be relatively more suitable as input for GOCI-II atmospheric correction.

Comparison of AMI and GEMS TCO for GOCI-II Atmospheric Correction







Poster #: 47

Date: Monday

Session #: 1

Title: Evaluation of Total Column Ozone from AMI and GEMS for

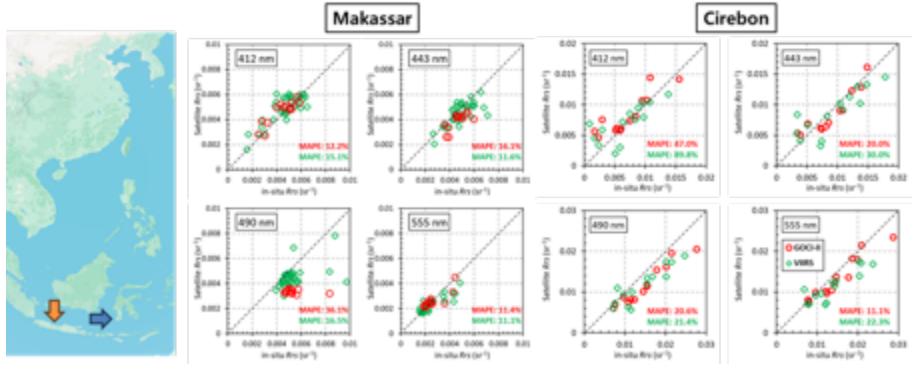
Atmospheric Correction of GOCI-II

[→] The AMI data is considered to be relatively more suitable as AC input material.

Status of GOCI-II atmospheric correction and Cal/Val



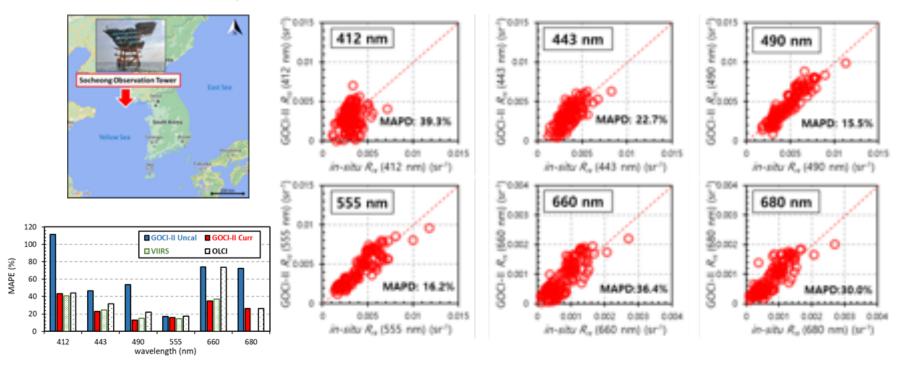
Match-ups from Indonesian Sea (Cirebon and Makassar)



Status of GOCI-II atmospheric correction and Cal/Val



Match-ups from Socheong Station (AERONET-OC)



Status of GOCI-II atmospheric correction and Cal/Val



•"

Match-ups from Socheong Station (AERONET-OC)

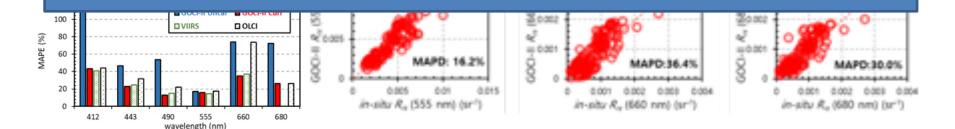
Poster #: 1

Date: Monday

Session #: 1

Title: Current Status of the GOCI-II Atmospheric Correction and Its

System Vicarious Calibration



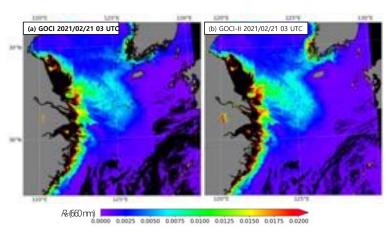
Comparison Rrs b/w GOCI & GOCI-II: East China Sea



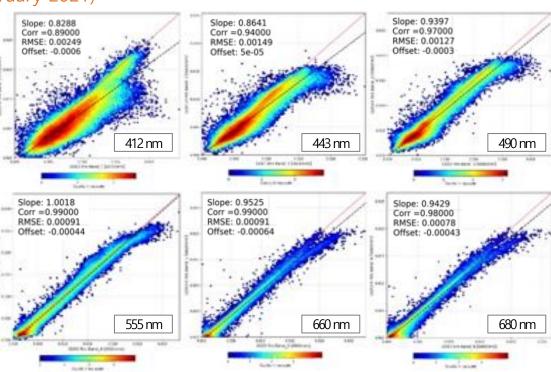
GOCI, GOCI-II (September 2020, February 2021)

Results of GOCI and GOCI-II by Spectral Band over the East China Sea (ECS)

	412 nm	443 nm	490 nm	555 nm	660 nm	680 nm
Corr.	0.89	0.94	0.97	0.99	0.99	0.98
RMSE	0.00249	0.00149	0.00127	0.00091	0.00091	0.00078



Rrs 660 nm images from GOCI(a), GOCI-II (b): (a, b) 21 February 2021. GOCI and GOCI-II 03 UTC images.

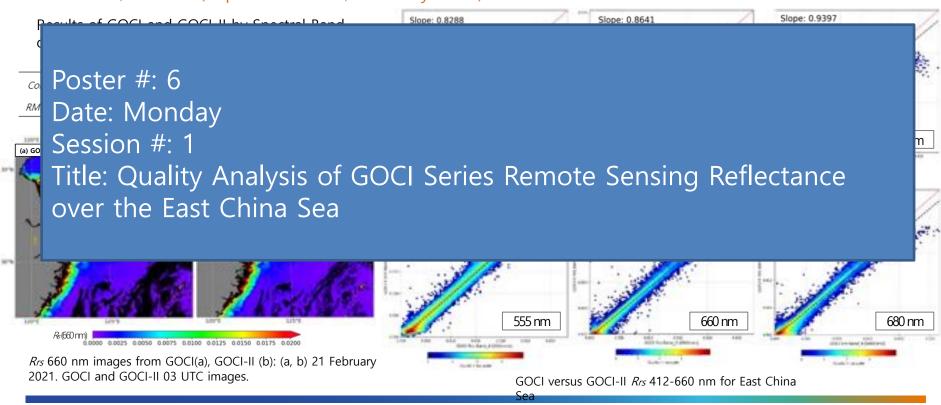


GOCI versus GOCI-II Rrs 412-660 nm for East China

Comparison Rrs b/w GOCI & GOCI-II: East China Sea



GOCI, GOCI-II (September 2020, February 2021)



Improvement of GOCI-II IOP





Hybrid IOP estimation using QAA and machine learning

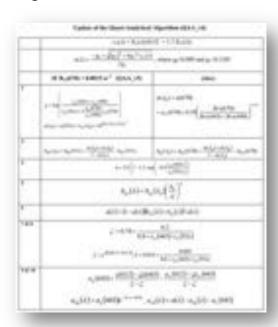
Estimated IOP using a radiative transfer simulation dataset (Hydrolight) based on machine learning and QAA

Туре	Brown Earth, Red Clay, Yellow Clay			
CDOM	Chl-a	TSM		
0.01 - 1.6654 m ⁻¹	0.1 - 17.7438 mg/m³	0.1 - 100 g/m ³		

	<i>J</i> ,
Input (11)	 Simulated U value of GOCI-II 7 bands (412, 443, 490, 510, 555, 620, and 660 nm) Bands ratio
Target (3)	 Backscatter coefficient of particles (b_{bp}) slope b_{bp} (555 nm) Absorption coefficient of phytoplankton (a_{ph}) (443 nm)
Machine learning models	 Artificial Neural Network (ANN) Kernel Ridge Regression (KRR) Histogram-based Gradient Boosting Regression (HGBR) Light Gradient Boosting Machine (LGBM)

Categorical Boosting (CatBoost)

- 1. Compute u from GOCI-II R_{rs}
- 2. Estimate $b_{bp}(555nm)$ and b_{bp} slope using the ML model
- 3. Compute spectral $b_{bp}(\lambda)$ using QAA
- 4. Compute spectral $b_b(\lambda)$ using QAA
- 5. Compute spectral $a(\lambda)$ using QAA
- 6. Estimate a_{ph}(443 nm) using the ML model
- 7. Compute $a_{dq}(443 \text{ nm})$ using QAA



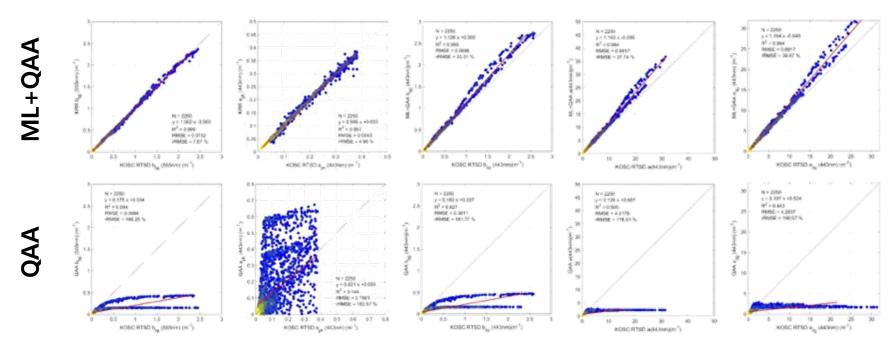
Improvement of GOCI-II IOP





Hybrid IOP estimation using QAA and machine learnning

Estimated IOP using a radiative transfer simulation dataset (Hydrolight) based on machine learning and QAA



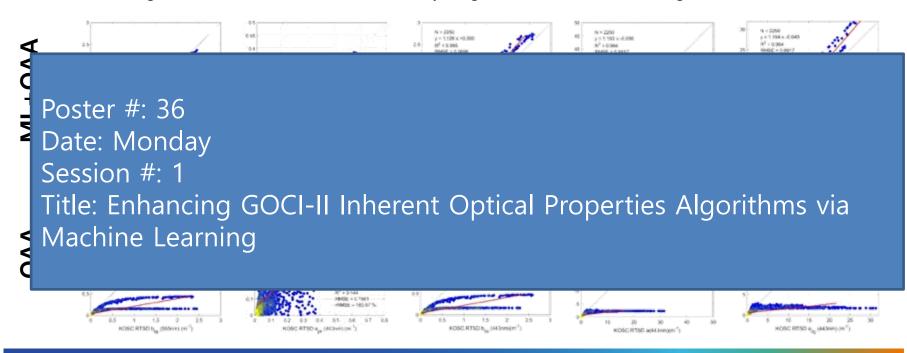
Improvement of GOCI-II IOP





Hybrid IOP estimation using QAA and machine learnning

Estimated IOP using a radiative transfer simulation dataset (Hydrolight) based on machine learning and QAA

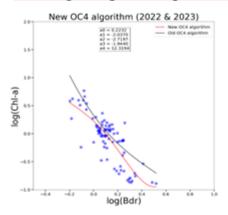


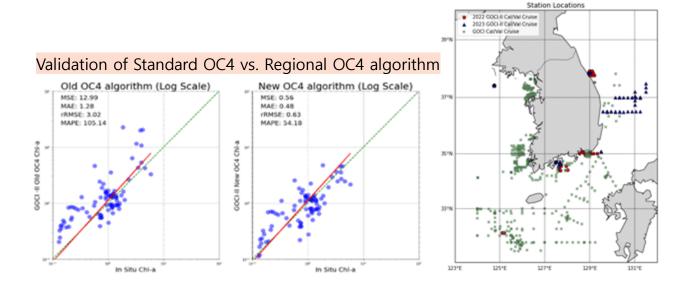
Updating GOCI-II regional Chl-a algorithm





Deriving a regional algorithm





- Development of the OC4 algorithm for the GOCI-II sensor in Korean coastal waters
- Adjustment of OC4 coefficients using the latest in-situ observations
- Validation of the new algorithm (showing improved accuracy)

Updating GOCI-II regional Chl-a algorithm

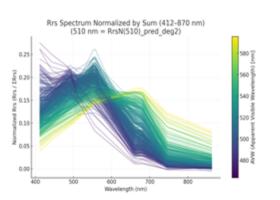




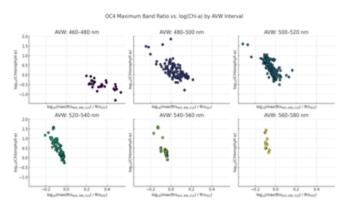
Development OC4 algorithm depending on water quality



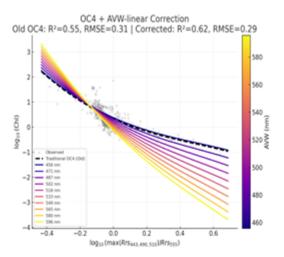
Global distribution of Apparent Visible Wavelength(AVW)



Band ratio – Chl-a relationship depending on water quality



OC4+water quality algorithm



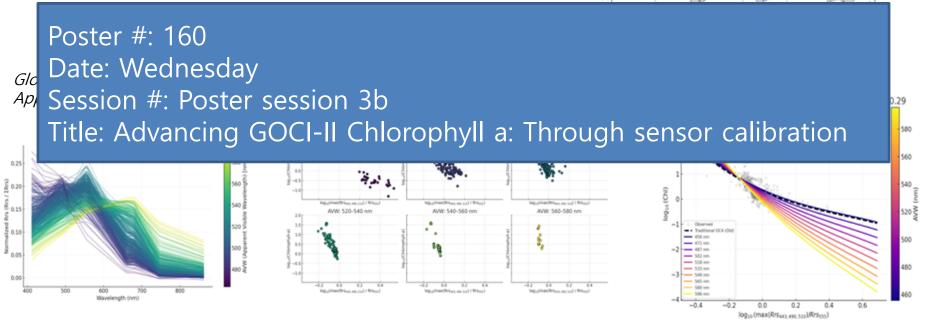
Updating GOCI-II regional Chl-a algorithm





Development OC4 algorithm depending on water quality



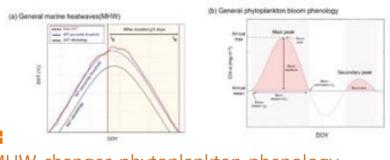


Marine Heatwave reshape phytoplankton phenology

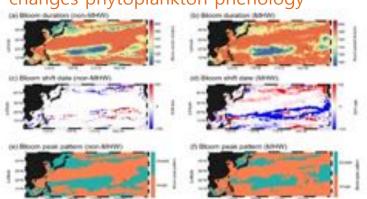




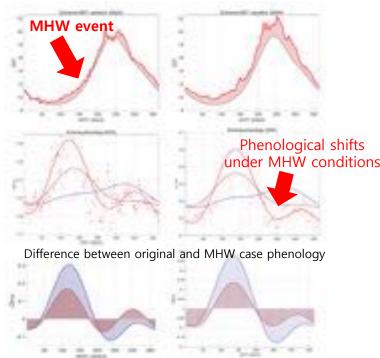
Defined marine heatwave(MHW) and phytoplankton phenology



MHW changes phytoplankton phenology



Please refer to **Poster No.##** for more details!



Marine Heatwave reshape phytoplankton phenology





Defined marine heatwave(MHW) and phytoplankton phenology

(b) General phytoplankton bloom phenology

Please refer to **Poster No.##** for more details!

Poster #: 7

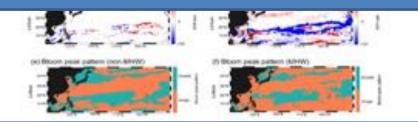
(a) General marine heatwaves(MHW)

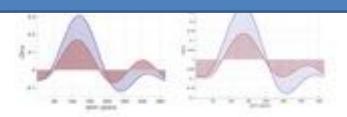
Date: Monday

Session #: 1a

Title: Marine Heatwaves Drive Shifts in Phytoplankton Phenology and

Seasonal Cycles





ifts itions

H/W specifications of GOCI series

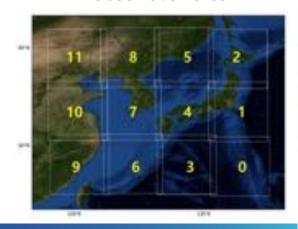


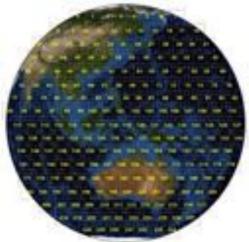
Spectral bands

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- Spatial resolution
 - GOCI
 - 500 m/pxl at projection center (36°N)
 (~400 m/pxl at nadir)
 - GOCI-II & GOCI-III
 - 250 m/pxl at nadir

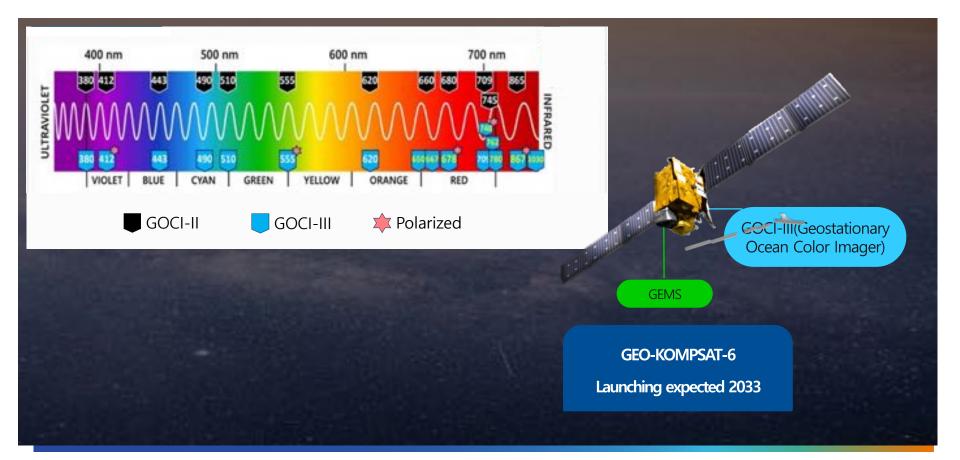
Observation area





Plan for GEO-KOMPSAT-6





Summary





The project on Cal/Val and algorithm improvement for GOCI-II will finish next year, and we are preparing the next project and trying to secure the budget to carry out the project until the end of GCI-II mission

We are developing and improving algorithms for radiometric calibration, atmospheric correction, IOP, Chlorophyll-a concentration, etc. to secure the accuracies of L2 until the end of / after the mission



Using nearly 20 years of GOCI-series data, we are trying to carry out diverse studies on seeing long term trend relating to climate changes like phenology study



We are preparing the development of GOCI-III with improved spectral design including polarization, which can be launched about 2023

Remote Sensing of Environment (RSE) Journal Special Issue



- Geostationary Ocean Color Remote Sensing Data for Improved Understanding of Surface Ocean Biogeochemistry
- Guest Editors
 - Dr. Jongkuk Choi, Korea Institute of Ocean Science & Technology (KIOST)
 - Dr. Myungsook Park, Korea Institute of Ocean Science & Technology (KIOST)
 - Prof. Joseph Salisbury, (PI of GLIMR), University of New Hampshire
 - Dr. Cara Wilson, (Program Leader of CoastWatch West Coast Node and of PolarWatch), NOAA
 - Dr. Hayley Evers-King, EUMESAT
- Submission deadline: 2026. 7. 31
- Key words: Geostationary Ocean Color, GOCI, GLIMR, GeoXO, SEVIRI
- Opinion from RSE Editors-in-chief
 - papers from many sensors GOCI, GOCI 2, Himawari 8, ABI and perhaps other Ocean GEO Sensors that can get meaningful information from the oceans (E.g. TEMPO).
 - set the target for at least 15 papers, but 20 is better. Also, while the rejection rate is typically 20% for RSE, it is often much higher (perhaps 50%) for special issue papers. So, we should shoot for at least 30 submissions

Remote Sensing of Environment (RSE) Journal Special Issue



• https://www.sciencedirect.com/special-issue/318780/geostationary-ocean-color-remote-sensing-data-for-improved-understanding-of-surface-ocean-biogeochemistry





