

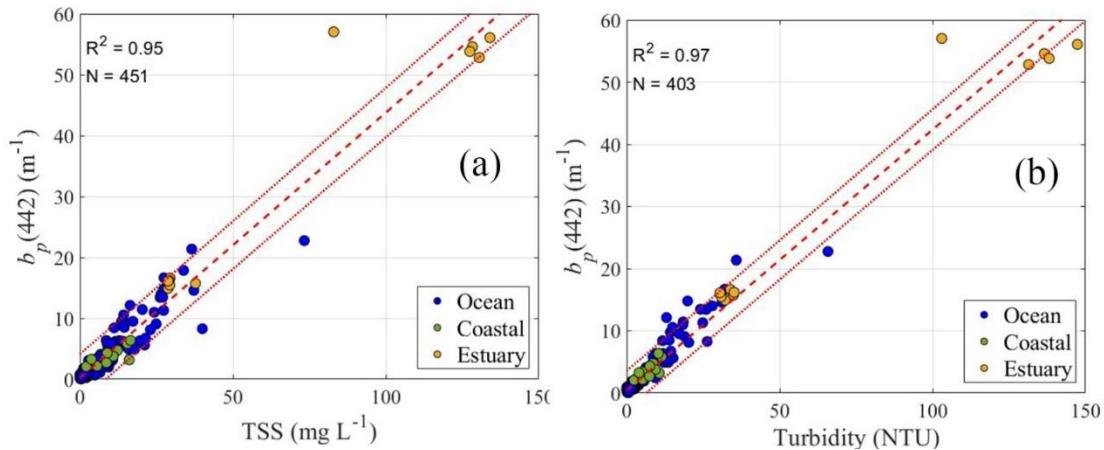
Supplement of:

# Assessment of Remote Sensing Reflectance Glint Correction Methods from Fixed Automated Above-Water Hyperspectral Radiometric Measurement in Highly Turbid Coastal Waters

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## S1- Estimation of $b_{bp}(\lambda)$ from $b_p(\lambda)$

In this study, we used the measured  $b_p(\lambda)$  and  $b_{bp}(\lambda)$  dataset by Röttgers et al. (2023) in the Wadden Sea (<https://doi.pangaea.de/10.1594/PANGAEA.954981>). The Total Suspended Sediment (TSS) and Turbidity showed strong correlation with  $b_p(\lambda)$  in oceanic (North Sea), coastal and estuaries regions of the Wadden Sea (Fig. S1). About 94% of sample data lied in the 95% confidence limit of variations. This indicates that the material scattering varied in a systematic way with increasing turbidity. Therefore, a trend of  $b_p(\lambda)$  from oceanic, coastal, and estuary indicated the scattering phase and variability of the particle backscattering ratio.



**Fig. S1-** Correlation between TSS (a) and Turbidity and  $b_p(442)$  using in-situ data. The dashed and dotted lines show the linear fits and the upper and lower bounds of 95% confidence level, respectively.

The  $b_{bp}(\lambda)$  of water components are determined by:

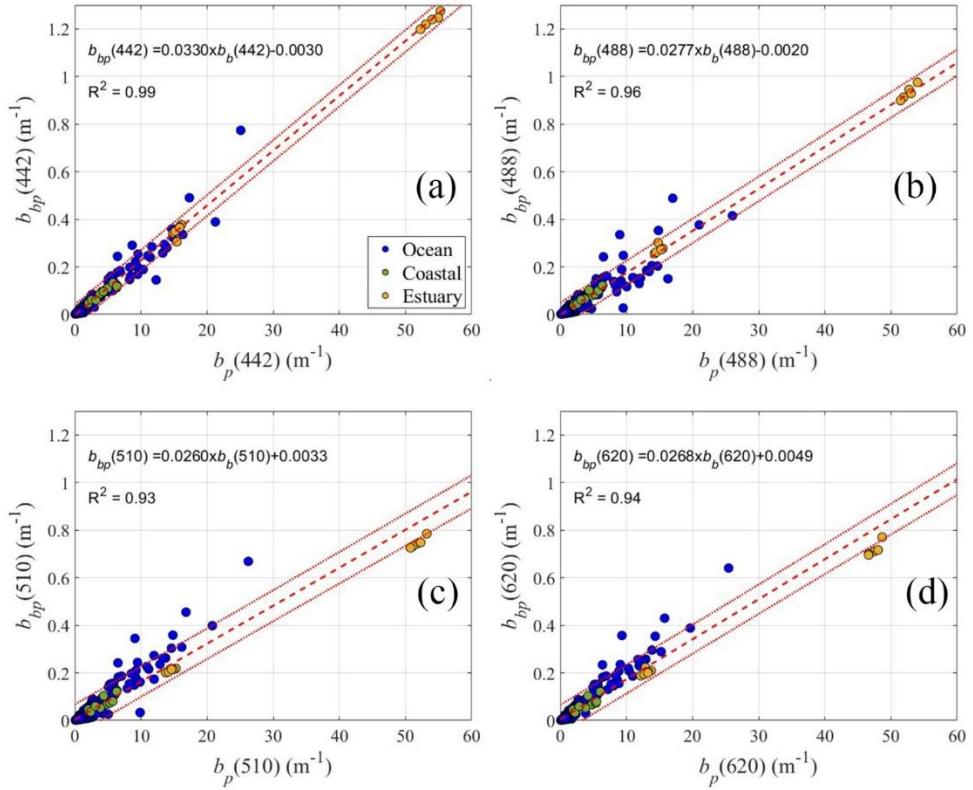
$$b_{bp}(\lambda) = b_p(\lambda) \times B \quad (S1)$$

where  $B$  is the backscattering ratio. Two approaches were considered to analyze the variability of  $B$  values (McKee et al., 2009), include: (i) point-by-point where  $B$  was calculated by dividing

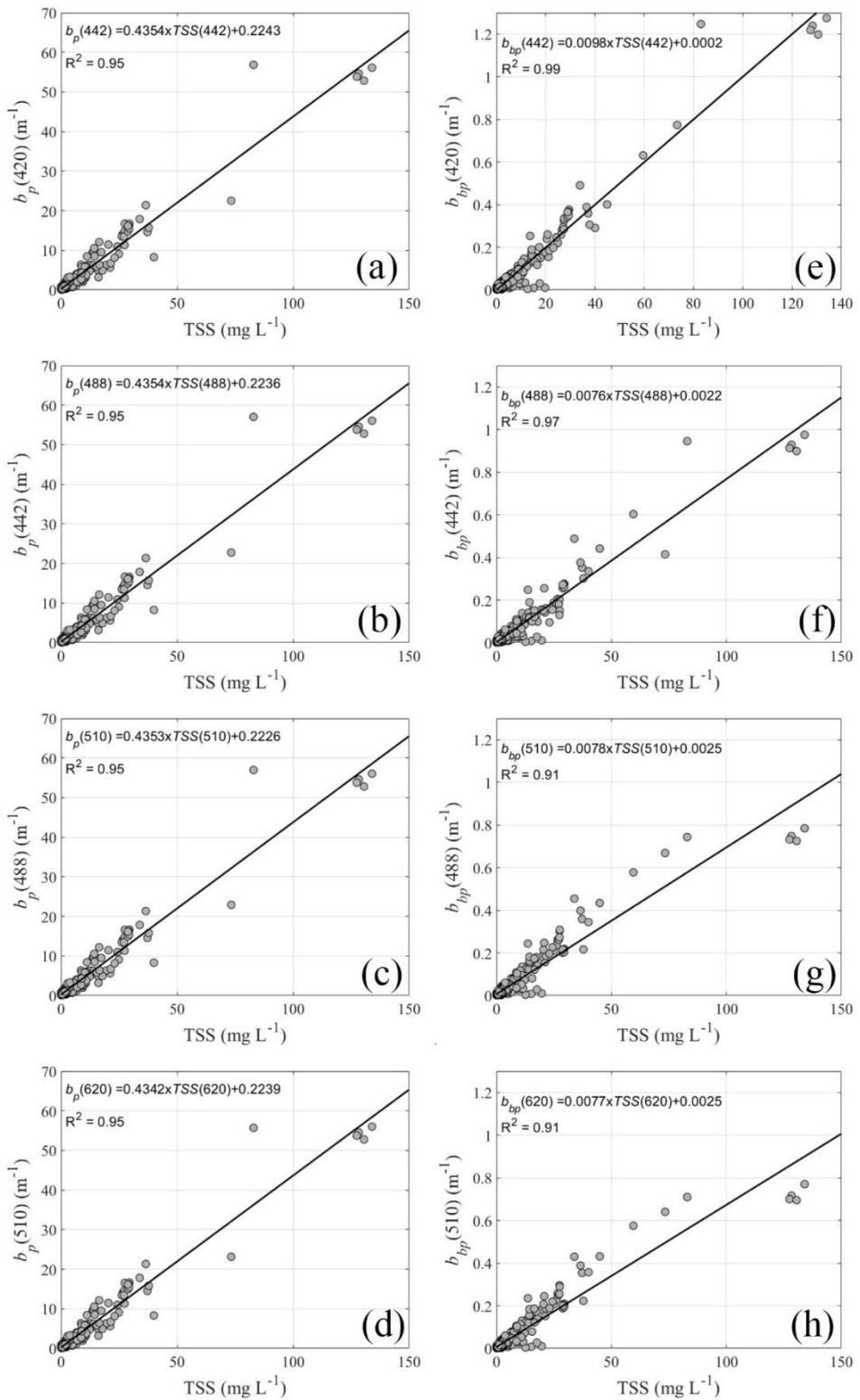
$b_{bp}(\lambda)$  by the corresponding individual  $b_p(\lambda)$  measurements. The variability of  $\mathbf{B}$  was analyzed using statistical metrics of the particle backscattering ratio distribution. (ii) regression analysis where the best fit values of  $b_{bp}(\lambda)/b_p(\lambda)$  for the whole measured dataset was calculated. To minimize the measurement uncertainties for both  $b_{bp}(\lambda)$  and  $b_p(\lambda)$ , the geometric mean regression model, known as the “Least Squares Bisector” (LSB) for Model II regression algorithm was used to analyze the spectral variability of  $\mathbf{B}$  values (<https://www.mbari.org/technology/matlab-scripts/linear-regressions/>). Since the bands at  $\lambda \leq 440$  nm and  $\lambda \geq 700$  nm introduced significant uncertainties in the spectral variability analysis, they were excluded from our analysis (Zhang et al., 2010).

Fig. S2 shows the correlation between measured  $b_{bp}(\lambda)$  and  $b_p(\lambda)$  at different wavelengths. About 92% of dataset lied in the 95% confidence level of  $b_{bp}(\lambda)$  and  $b_p(\lambda)$  variations. The outlier data may be due to variations in the material composition that change with increasing turbidity, and possibly also due to the performance limitations of field measurements. A linear relationship was found in Fig. S2 for all wavelengths. Hereafter, the results were presented for dataset which lie in the 95% confidence level of  $b_{bp}(\lambda)$  vs.  $b_p(\lambda)$ . The value of  $b_{bp}(\lambda)/b_p(\lambda)$  values in Fig. S3 were 0.0330, 0.0277, 0.0260, and 0.0268 with 95% confidence intervals of  $\pm 0.0002$  for bands 442, 488, 510, and 620, respectively. The maximum relative error of 30.4% was observed between bands 420nm and 510nm. The offsets in Fig. S2 were unexpected, as the zero values of  $b_p(\lambda)$  and  $b_{bp}(\lambda)$  should have coincided (Eq. S1). This could be due to an underestimation of  $b_{bp}$ , an overestimation of  $b_p$ , or a combination of both. Fig. S3 shows the correlations of TSS with  $b_p(\lambda)$  and  $b_{bp}(\lambda)$ . The positive offset of the linear fits in the correlation between TSS and  $b_p(\lambda)$  revealed underestimation of TSS or overestimation of  $b_p$  ( $\Delta b_p$ ). Meanwhile, the positive offset between TSS and  $b_{bp}(\lambda)$  showed underestimation of TSS or overestimation of  $b_{bp}$  ( $\Delta b_{bp}$ ). Therefore, the non-zero offset in Fig. S2 resulted possibly from overestimation of  $b_p$  and  $b_{bp}$ .

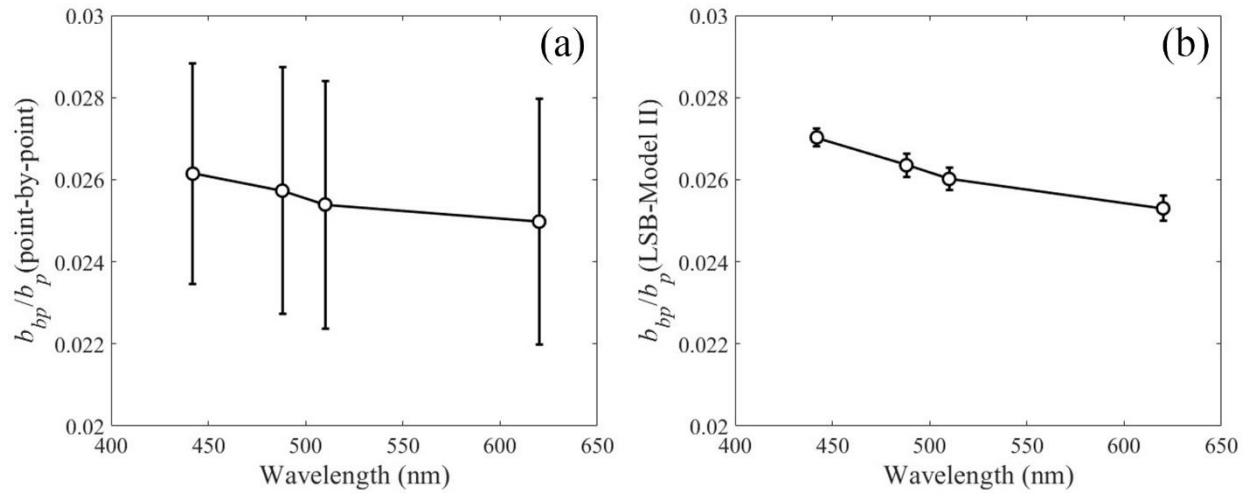
The values of  $\Delta b_p$  and  $\Delta b_{bp}$  were calculated using the geometric mean regression LSB-Model II from dataset of Fig. S3a-d and Fig. 3f-h, respectively. The values of  $\Delta b_p$  were 0.1592, 0.1585, 0.1579, and 0.1617 for bands 442, 488, 510, and 620 nm, respectively. The values of  $\Delta b_{bp}$  were 0.0016, 0.0009, 0.0021, and 0.0022 for bands 442, 488, 510, and 620 nm, respectively. Fig. 4S shows the mean values of backscattering ratio,  $B$ , at different wavelengths after correction of  $\Delta b_p$  and  $\Delta b_{bp}$ . The mean values of point-by-point  $B$  were  $0.0256 \pm 0.0027$  at wavelengths 442-620 nm (Fig. S4a). Fig. S4b shows the LBS-Model II regression of backscattering ratio at different wavelengths and associated 95% confidence intervals of  $\mathbf{B}$ . The LSB-Model II regression determined a narrow confidence intervals of backscattering ratio, with a mean value of  $0.0261 \pm 0.0002$ . As a result, the backscattering ratio could be regarded as wavelength-independent due to the measurements associated with  $b_p(\lambda)$  and  $b_{bp}(\lambda)$  in the range of 442-620 nm.



**Fig. S2-** Correlation between  $b_p$  and  $b_{bp}$  at 442 nm (a), 488 nm (b), 510 nm (c), and 620 nm (d). The dashed and dotted lines show the linear fits and the upper and lower bounds of 95% confidence level, respectively.  $N = 483$ .

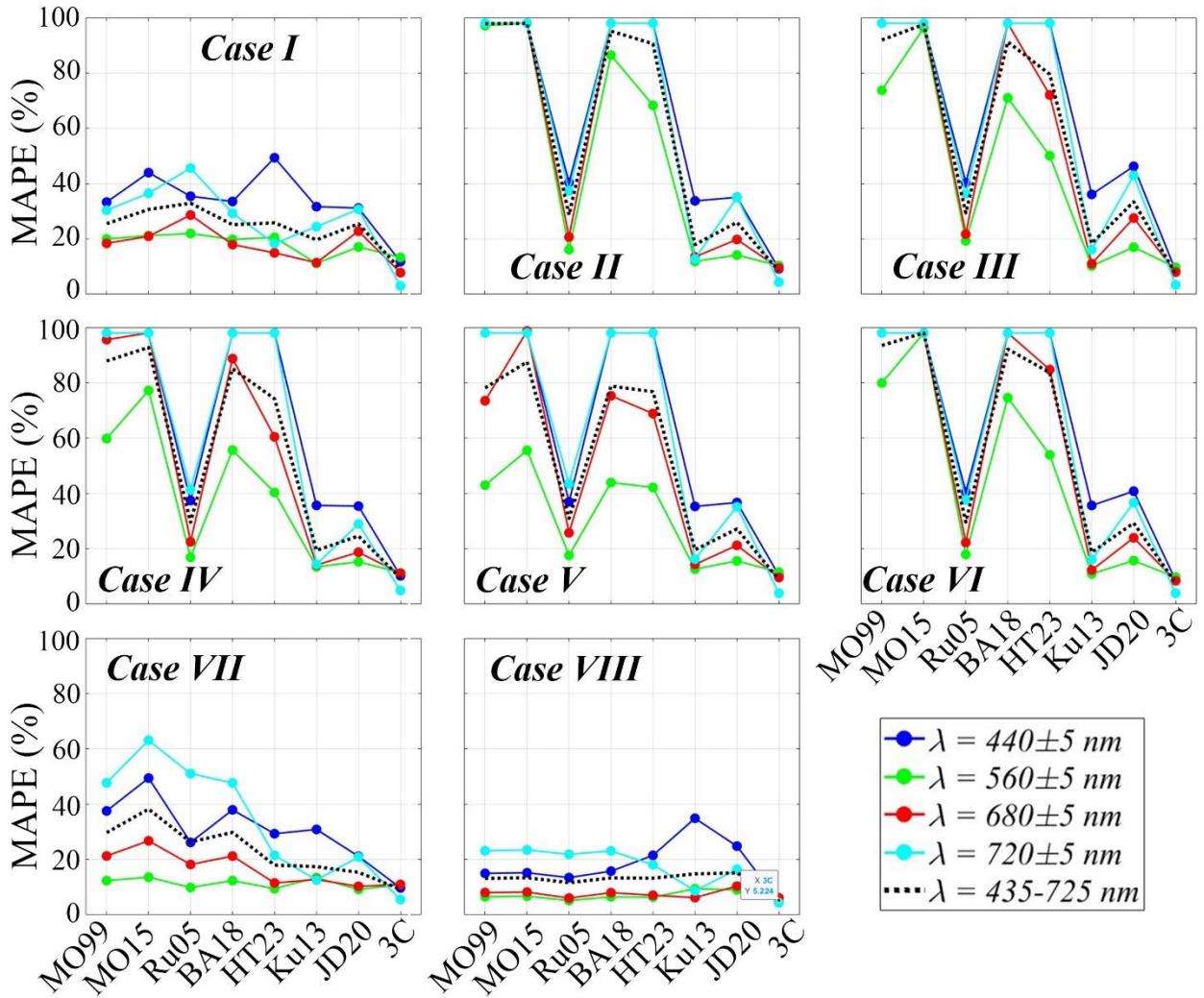


**Fig. S3-** Correlation between TSS and  $b_p$  (left column) and  $b_{bp}$  (right column). The solid lines show the best linear fits. N = 451.

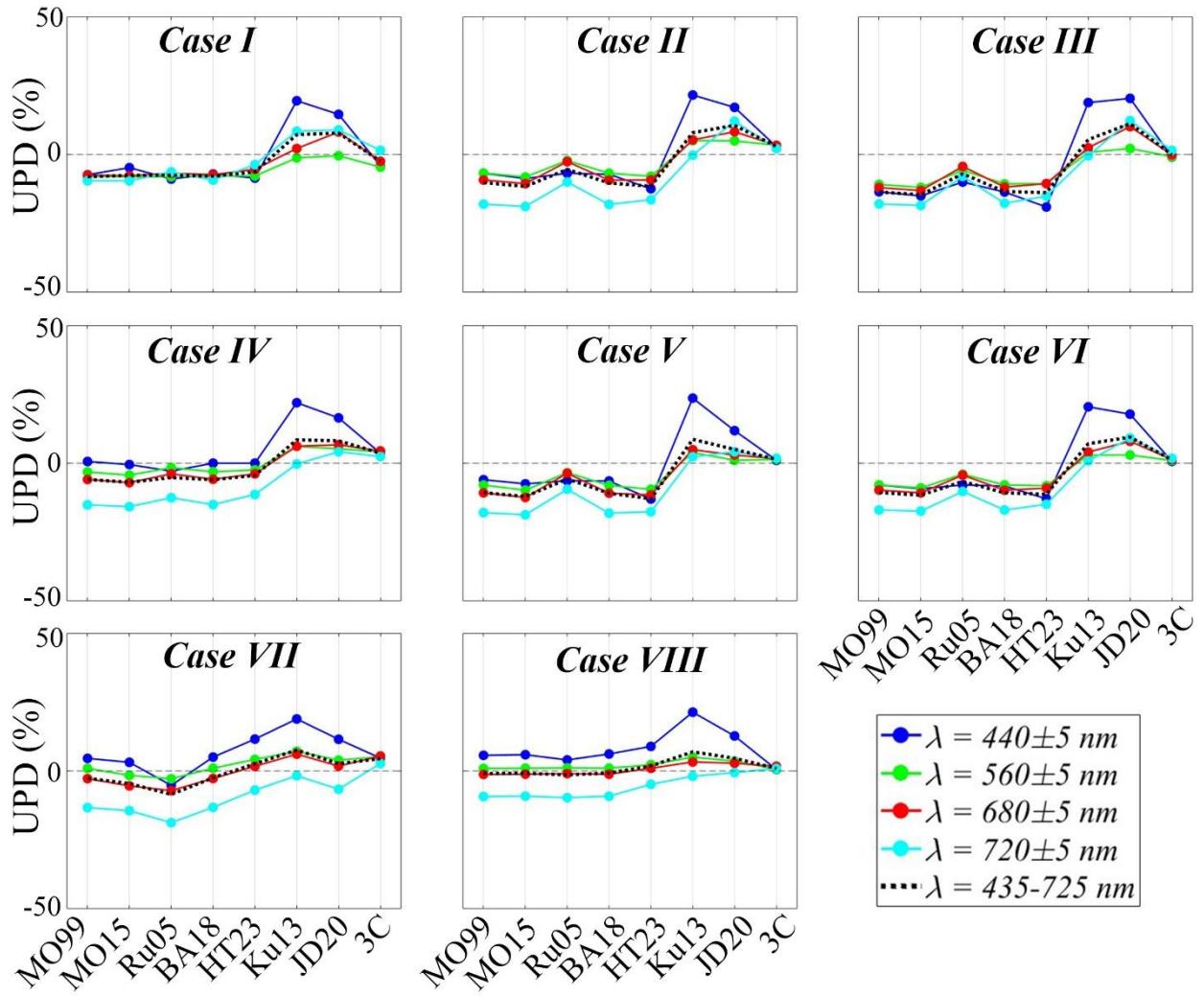


**Fig. S4-** **a)** variation of backscattering ratio at different wavelength using the point-by-point method. Error bars show  $\pm 1$  standard deviation. **b)** similar to (a), but calculated using the LSB-Model II regression method. Error bars show tight confidence intervals.

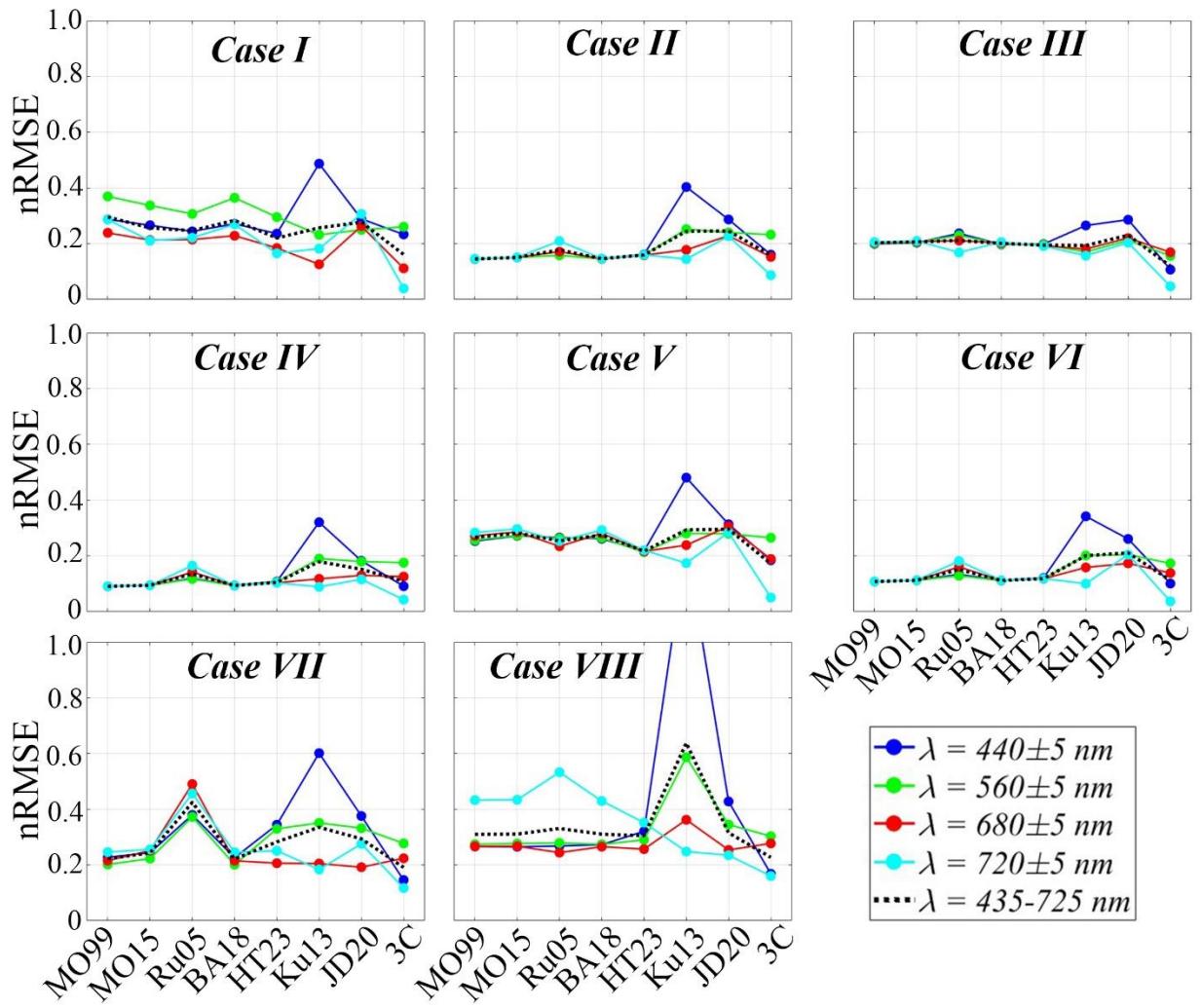
## S2- Statistical Parameters



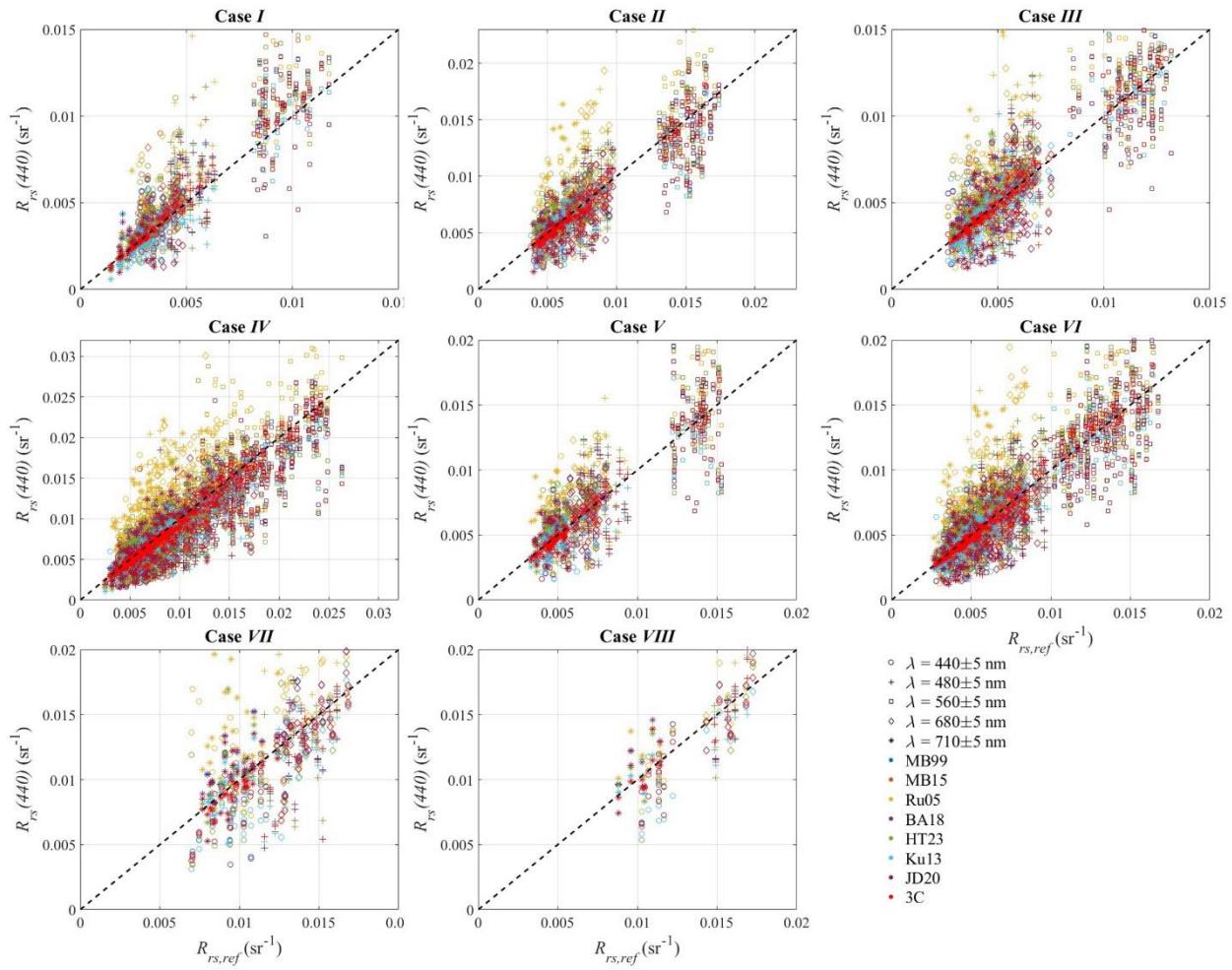
**Fig. S5.** The MAPE between  $R_{rs,ref}(\lambda)$  and modeled  $R_{rs}(\lambda)$  in different environmental conditions (Case I - Case VIII). The environmental conditions are illustrated in the text.



**Fig. S6.** The UPD values between  $R_{rs,ref}(\lambda)$  and modeled  $R_{rs}(\lambda)$  in different environmental conditions (Case I - Case VIII). The environmental conditions are illustrated in the text.



**Fig. S7.** The nRMSE values between  $R_{rs,ref}(\lambda)$  and modeled  $R_{rs}(\lambda)$  in different environmental conditions (Case I - Case VIII). The environmental conditions are illustrated in the text.



**Fig. S8.** Scatterplots of  $R_{rs,ref}(\lambda)$  vs. estimated  $R_{rs}(\lambda)$  using different models at selected blue, green, red, and NIR wavelengths in different environmental conditions (Case I - Case VIII). The environmental conditions are illustrated in the text.

## Reference:

Röttgers, R., Bi, S., Burmester, H., Heymann, K., Hieronymi, M., Krasemann, H., Schönfeld, W., 2023. Water inherent optical properties and concentrations of water constituents from the German Bight and adjacent regions. <https://doi.org/10.1594/PANGAEA.950774>