

Spectral Classification of MODIS and VIIRS Water-Leaving Reflectance

Ryan Vandermeulen, Antonio Mannino, Susanne Craig, Jeremy Werdell Ocean Ecology Laboratory NASA Goddard Space Flight Center

> MODIS/VIIRS Science Team Meeting College Park, MD November 19, 2019



Optical "Fingerprints"



How can quantify and analyze the differences between multiple layers of spectral information within the context of a two-dimensional image?



Spectral Classification (abridged history)

- Fuzzy c-means classification (Moore et al. 2009, 2014, Eleveld et al. 2017)
- Agglomerative Ward's linkage (Lubac and Loisel 2007)
- Iterative self-organizing data analysis technique (Melin and Vantrepotte 2015)
- Varimax-rotated Principal Component Analysis (Avouris and Ortiz 2019)
- Max-classification (Ye et al. 2016)
- K-means clustering (Wei et al. 2016, Prasad and Agarwal 2016)
- Chromaticity coordinates/Hue angle CIE (Wernand et al. 2013)

PITFALLS

- Usually require training data sets (hard to come by)
- Often yield regionally-specific results, requires development
- Arbitrarily define number of clusters in some instances
- Exist in dimensionless space/not quantitative (Class A, B, C, etc.)
- May not incorporate the entirety of the $R_{rs}(\lambda)$ spectrum
- Challenging to interpret without some a priori knowledge of the data



Wavelength (nm)

Apparent Visible Wavelength

$$AVW = \frac{\sum_{i=\lambda_{1}}^{\lambda_{n}} R_{RS}\left(\lambda_{i}\right)}{\sum_{i=\lambda_{1}}^{\lambda_{n}} \frac{R_{RS}\left(\lambda_{i}\right)}{\lambda_{i}}} = \left(\frac{\sum_{i=\lambda_{1}}^{\lambda_{n}} \lambda_{i}^{-1} R_{RS}\left(\lambda_{i}\right)}{\sum_{i=\lambda_{1}}^{\lambda_{n}} R_{RS}\left(\lambda_{i}\right)}\right)^{-1}$$

The weighted harmonic mean of the R_{RS} wavelengths, outputs an **Apparent Visible Wavelength, AVW** (in units of nm). The derivation of the AVW is simply a first-order measure of the dominant color of the water, as determined by the weight that each measured channel contributes to the reflectance in the visible range of the spectrum.



Apparent Visible Wavelength



Every color on the map represents a different spectral shape. Constraining multiple layers (wavelengths) of multi-spectral $R_{rs}(\lambda)$ into one dimension is a simple and robust way for users to visualize and quantify trends in spectral $R_{rs}(\lambda)$ in terms of its apparent dominant color, which, inherently relates to a specific spectral shape and a unique combination of absorption and scattering properties.





IOCCG Synthetic Dataset sorted according to Apparent Visible Wavelength



Similar $R_{rs}(\lambda)$ spectral shapes tend to converge along the AVW gradient, ultimately constrained by the finite amount of combinations of absorption and backscatter found in the natural world that create a particular $R_{rs}(\lambda)$ spectrum with an identical balance point.

VIIRS

Disparate sensors sense the Earth differently

MODIS

OCTS

PACE



SeaWiFS

MERIS

Deriving continuity in spectral shape from heritage sensors of disparate resolutions

- 1) $R_{rs}(\lambda)$ from each heritage sensor (SeaWiFS, MODIS-Aqua, and VIIRS) is reconstructed from the 720 IOCCG hyperspectral $R_{rs}(\lambda)$, using the Relative Spectral Response (RSR) functions for each sensor.
- 2) The AVW is calculated for hyperspectral data (only 400-700 nm, for compatibility), and compared to AVW from each sensor.
- 3) 3rd or 4th polynomial fits are retained and applied to satellite data, enabling the analysis of comparable AVW values.







2012 – 2018 MODIS-Aqua v. SNPP-VIIRS differences in spectral shape

AVW difference (nm), date: 2012001



A monthly time series of the differences in spectral shape (AVW) will episodically shift lower, manifested as a blue-shift in MODIS-Aqua $R_{rs}(\lambda)$ relative to SNPP-VIIRS.



Elucidating spectral shifts over time (SeaWiFS \rightarrow MODIS/VIIRS \rightarrow PACE)





Robust Linear regression/time of AVW enables the examination of spectral shift in $R_{rs}(\lambda)$ over time

Applications



- Identification of optical water types (OWT), correlation of similar water types on global scales
- Global spectral matching function to analyze trends between disparate sensors.
- Analysis of spectral variance for targeted sampling
- Potential improvements to parameterization for semianalytical inversions
- A useful climatological metric of change
- Useful for display of multi/hyperspectral in situ data
- Implementation of decision tree approaches for algorithm development
- Quality control check of algorithm performance (e.g. erratic spectral shapes)
- When used in tandem with other biophysical parameters, PFT and/or water-type distinction



Extras

Sanity Check → test on independent Hyperspectral dataset



A hypothetical AVW for MODIS is calculated from **HICO** data, by applying MODIS RSRs. This serves as an independent check on the quality of the polynomials. Since the values are vary similar, we should be able to directly compare an actual MODIS image to a HICO image to examine spectral shifts relative to one another.

Uncertainty in AVW to represent spectral shape can lead to Type II, but not Type I errors



One further step of classification by the wavelength of maximum reflectance (λ_{max}) can help alleviate Type II errors and further reduce uncertainty by a factor of 5.



AVW v.

Hue Angle



On average, each 1-nm increment of AVW has a mean absolute deviation of 3.09 hue angle degrees. The disparities between the two are likely a function of the RSR of the human eye not being sensitive to far blue/far red portions of the spectrum.



Hovmöller diagrams can lend enhanced insight into the spatiotemporal distribution of $R_{rs}(\lambda)$.



Potential Discards

The challenge:

When analyzing large satellite data sets, it is challenging to represent and comprehend more than two dimensions of data at one time.

Graphical representation of spectral data:

- Multiple $R_{RS}(\lambda)$ from **one** pixel/ROI plotted over time
- **One** $R_{RS}(\lambda)$ or product (chlor_a), mapped in 2 dimensions
- Δ time (animation), but only for **one** product/ λ at a time
- Multiple $R_{RS}(\lambda)$ scatter plot over region of interest
- *RGB image to "ID" coccolithophore blooms, qualitative*



How can quantify and analyze the differences between multiple layers of spectral information within the context of a two-dimensional image?



Wavelength (nm)