

# Evaluation of unoccupied aircraft system (UAS) remote sensing reflectance retrievals for water quality monitoring in coastal waters

Anna Windle

NASA Goddard Space Flight Center / Science Systems and Applications, Inc.



**FRM4DRONES**  
fiducial reference  
measurements  
for water using  
drones



# Outline

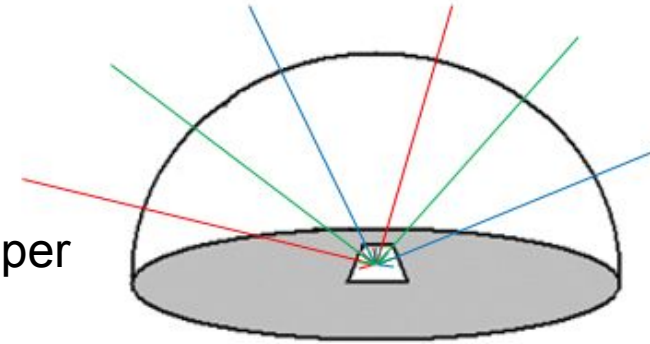
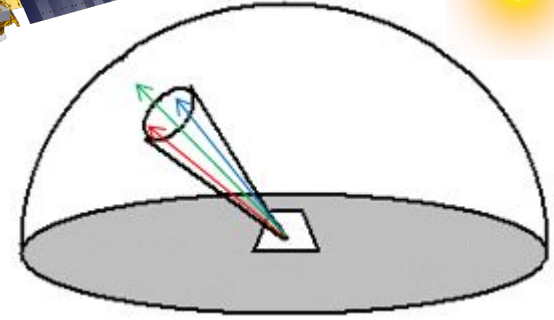
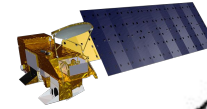
- Theory on above-water remote sensing
- Methods to remove surface reflected light
  - Windle & Silsbe, 2021
- DroneWQ Python package
- Future directions



# Radiometric measurements

- Remote sensors measure **radiance**: radiant flux per unit of solid angle
  - Typical units:  $\text{W m}^{-2} \text{sr}^{-1}$ , where sr is steradian, a measure of a solid angle
- Radiance is dependent on the illumination of observed area, or the hemispherical downwelling solar **irradiance**: radiant flux incident on a surface per unit area
  - Typical units:  $\text{W m}^{-2}$

$$\text{surface reflectance } (\text{sr}^{-1}) \approx \frac{\text{radiance}}{\text{irradiance}}$$

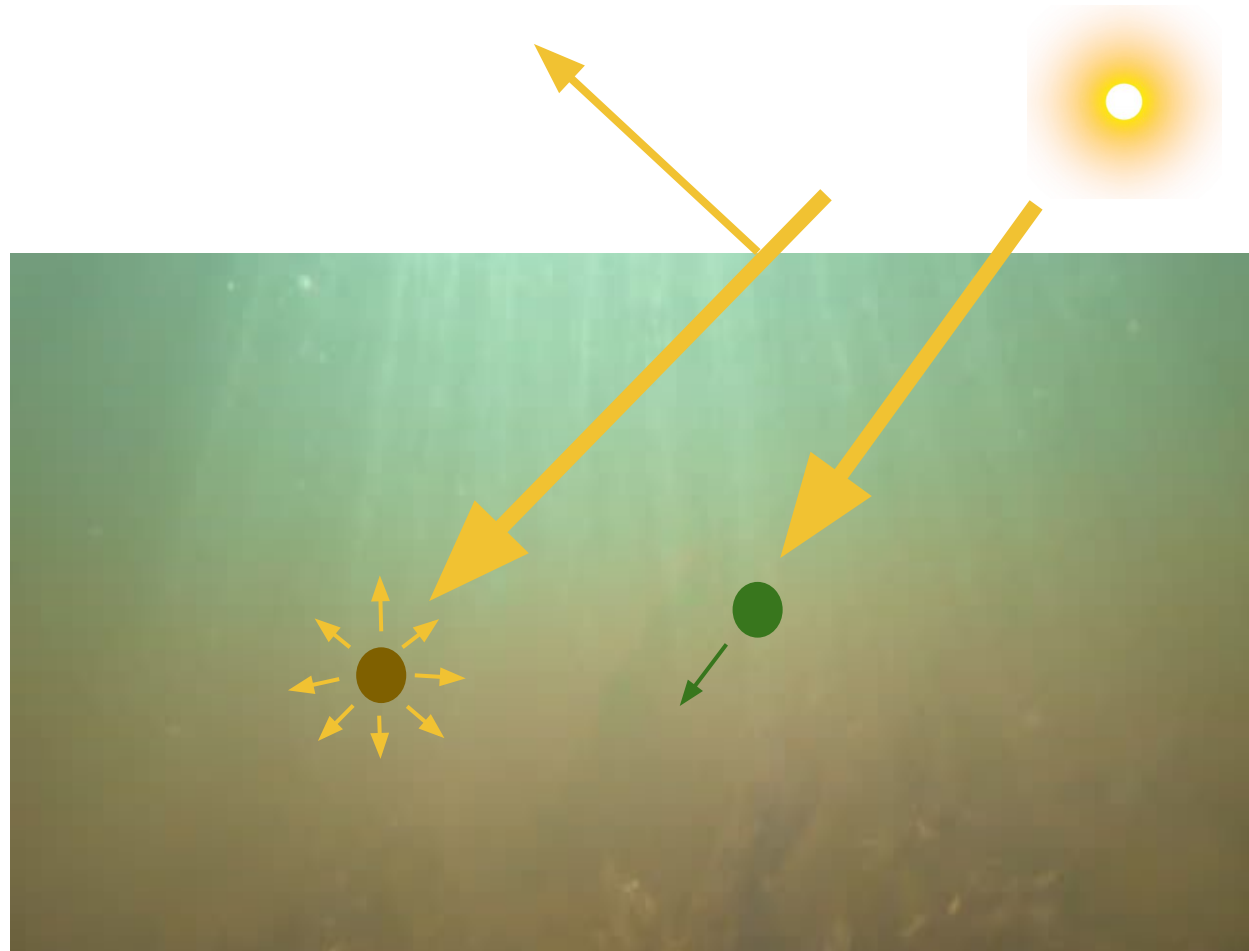


Credit: Anders Knudby,  
CC BY 4.0.

# Light in water

Two important processes determine fate of light in water:

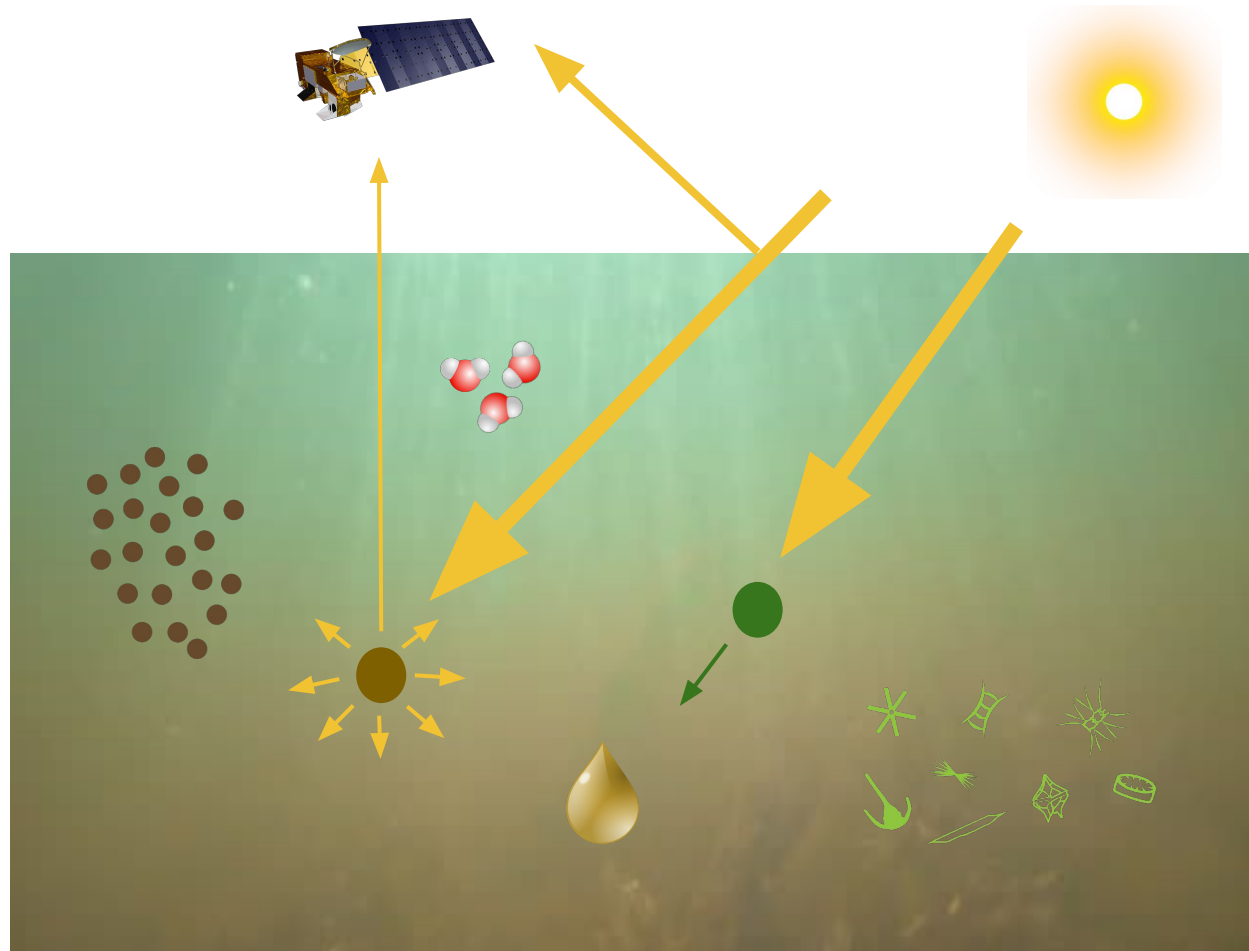
- 1) **Absorption:** light is taken up by molecules or optically active constituents
- 2) **Scattering:** light is redirected with no loss



# Light in water

Two important processes determine fate of light in water:

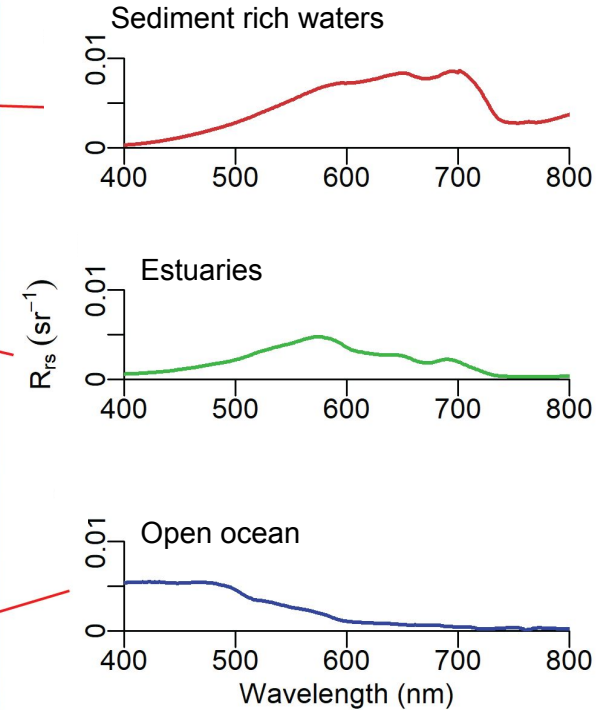
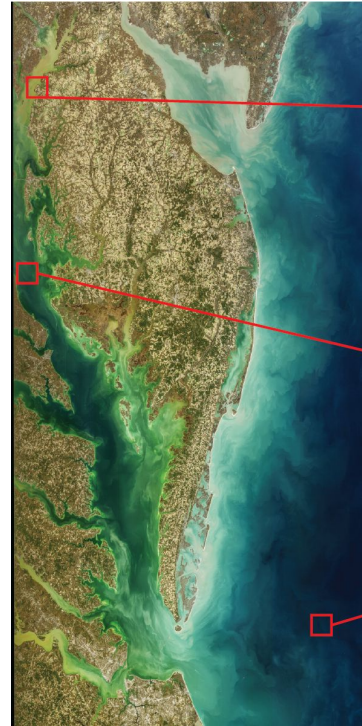
- 1) **Absorption:** light is taken up by molecules or water matter
- 2) **Scattering:** light is redirected with no loss





# Remote sensing reflectance ( $R_{rs}$ )

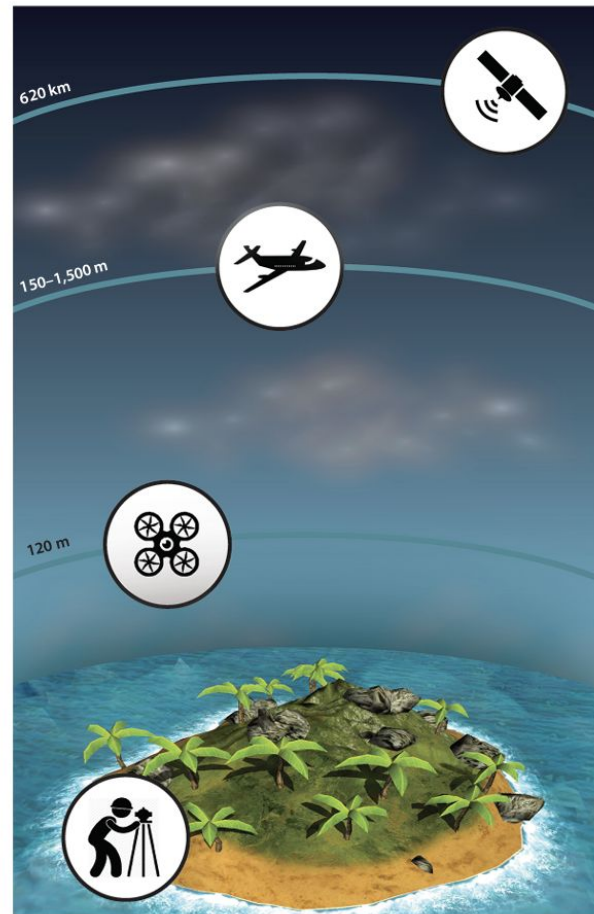
Absorption and backscattering of optically active constituents determines color of water -> can be used to infer what's in the water



# UAS fill a gap



Photo credit: Matt Pluta, ShoreRivers



# UAS seek to measure remote sensing reflectance ( $R_{rs}$ )

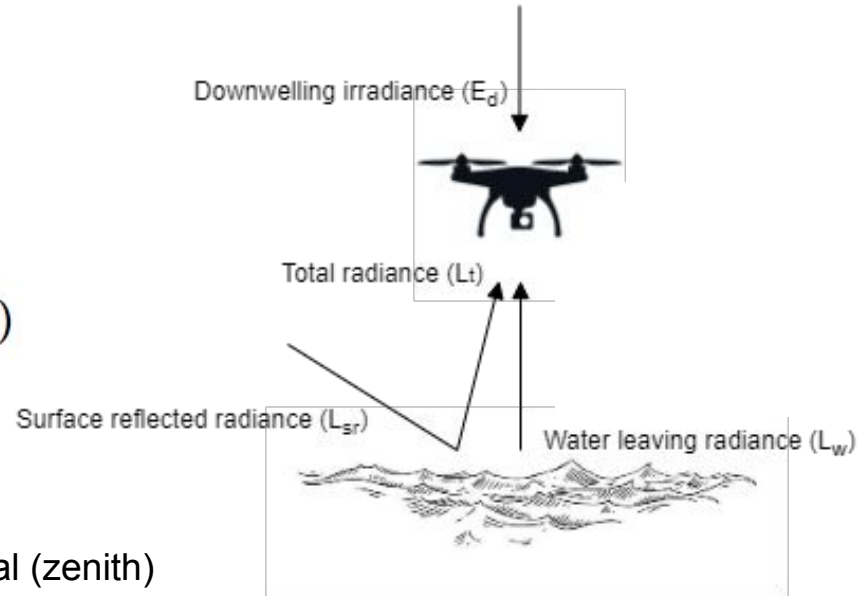
$$R_{rs}(\theta, \Phi, \lambda) = \frac{L_W(\theta, \Phi, \lambda)}{E_d(\lambda)}$$

$$L_T(\theta, \Phi, \lambda) = L_W(\theta, \Phi, \lambda) + L_{SR}(\theta, \Phi, \lambda)$$

$\theta$  is sensor viewing angle between the sun and the vertical (zenith)

$\phi$  is the angular direction relative to the sun (azimuth)

$\lambda$  represents wavelength





# Research motivation and question

**M:** UAS equipped with optical sensors offer on-demand, highly resolved data not amenable to satellite remote sensing. Aquatic UAS applications are in their infancy, and the removal of light reflected directly off the skin of water has received little attention in the literature.

***Q:*** *How can sun glint and surface reflected light be removed from UAS imagery to obtain accurate water leaving radiance?*

Reference	UAS sensor(s)	Radiometric quantity	Removal of $L_{SR}$ ?	WQ parameter(s)
Zeng et al. (2017)	Ocean optics STS-VIS spectrometers (hyperspectral)	$R_{UAS}$	No	Chl <i>a</i> , CDOM, turbidity
Shang Z. et al. (2017)	AvaSpec-dual spectroradiometers (hyperspectral)	$R_{rs}$	Yes	Chl <i>a</i>
Su, (2017)	Canon powershot S110 RGB and NIR sensors	$R_{UAS}$	Yes	Chl <i>a</i> , secchi disk depth, turbidity
Choo et al. (2018)	MicaSense RedEdge and DLS (multispectral)	$R_{UAS}$	No	Chl <i>a</i>
Baek et al. (2019)	MicaSense RedEdge and DLS (multispectral)	$R_{rs}$	Yes	Chl <i>a</i>
Becker et al. (2019)	Ocean optics STS-VIS spectrometers (hyperspectral)	$R_{UAS}$	No	Cyanobacteria index, chl <i>a</i> TSS
Arango and Nairn. (2019)	MicaSense RedEdge and DLS (multispectral)	$R_{UAS}$	No	Secchi disk depth, chl <i>a</i> , TSS, TN, TP
Olivetti et al. (2020)	Parrot sequoia (multispectral)	$R_{UAS}$	No	TSS
McEliece et al. (2020)	Sentera multispectral sensor (4 visible bands)	$R_{UAS}$	No	Chl <i>a</i> , turbidity
Kim et al. (2020)	MicaSense RedEdge-M and DLS (multispectral)	$R_{rs}$	Yes	Chl <i>a</i> (but not focus of paper)
Castro et al. (2020)	MicaSense RedEdge and DLS (multispectral)	$R_{rs}$	No	Chl <i>a</i>
O'Shea et al. (2020)	Resonon Pika L spectrometer (hyperspectral) *deployed on a tower, not UAS	$R_{rs}$	Yes	Chl <i>a</i>

Windle & Silsbe, 2021

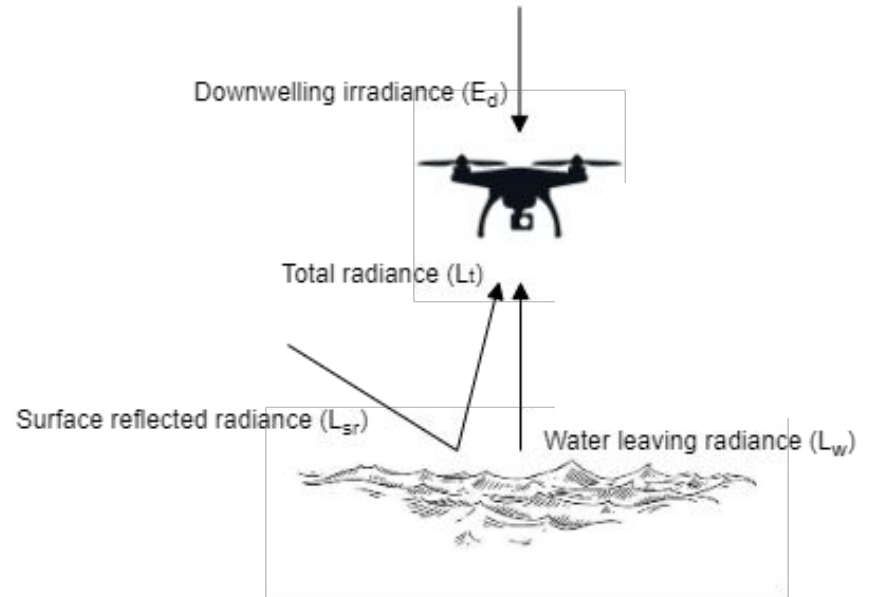
We want...

$$R_{rs}(\theta, \Phi, \lambda) = \frac{L_W(\theta, \Phi, \lambda)}{E_d(\lambda)}$$

But, a UAS measures  $L_T$ ...

$$L_T(\theta, \Phi, \lambda) = L_W(\theta, \Phi, \lambda) + L_{SR}(\theta, \Phi, \lambda)$$

How do we account for  $L_{SR}$ ?



# Modelling surface reflectance radiance ( $L_{SR}$ )

## Estimation of the remote-sensing reflectance from above-surface measurements

Curtis D. Mobley

APPLIED OPTICS / Vol. 38, No. 36 / 20 December 1999

*“The numerical simulations suggest that a viewing direction of 40 deg from the nadir and 135 deg from the Sun is a reasonable compromise among conflicting requirements. For this viewing direction, a value of  $\rho = 0.028$  is acceptable only for wind speeds less than  $5 \text{ m s}^{-1}$ ”*

$$L_{SR}(\theta, \Phi, \lambda) = \rho(\theta, \Phi, \lambda) * L_{sky}(\theta, \Phi, \lambda)$$

(effective sea surface  
reflectance of the  
wave facet =  
0.028) (Pic of sky)

Surface reflected light ( $L_{SR}$ ) can vary across pixels



Photo credit: Anna Windle



Photo credit: Patrick Gray

Surface reflected light ( $L_{SR}$ )



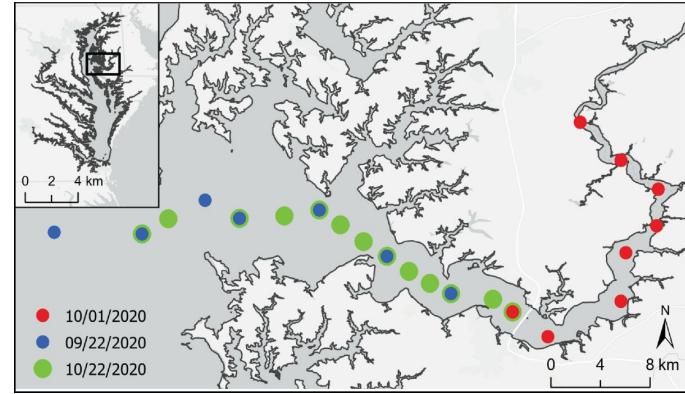


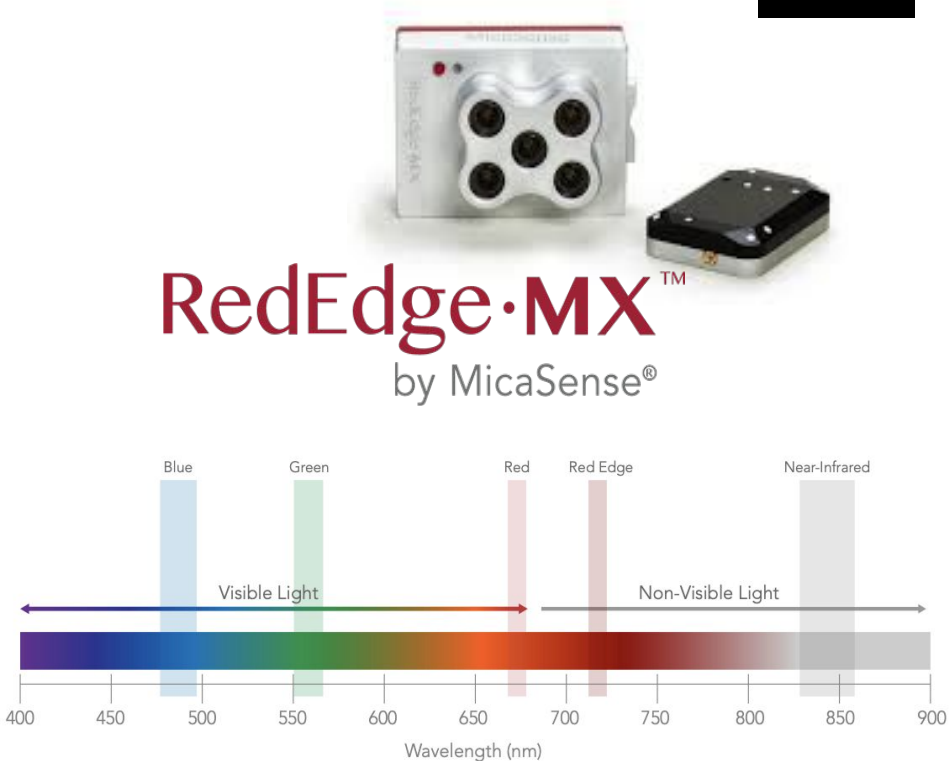
# Evaluation of Unoccupied Aircraft System (UAS) Remote Sensing Reflectance Retrievals for Water Quality Monitoring in Coastal Waters

*Anna E. Windle\* and Greg M. Silsbe*

*Horn Point Laboratory, University of Maryland Center for Environmental Science, Cambridge, MD, United States*

# Methods

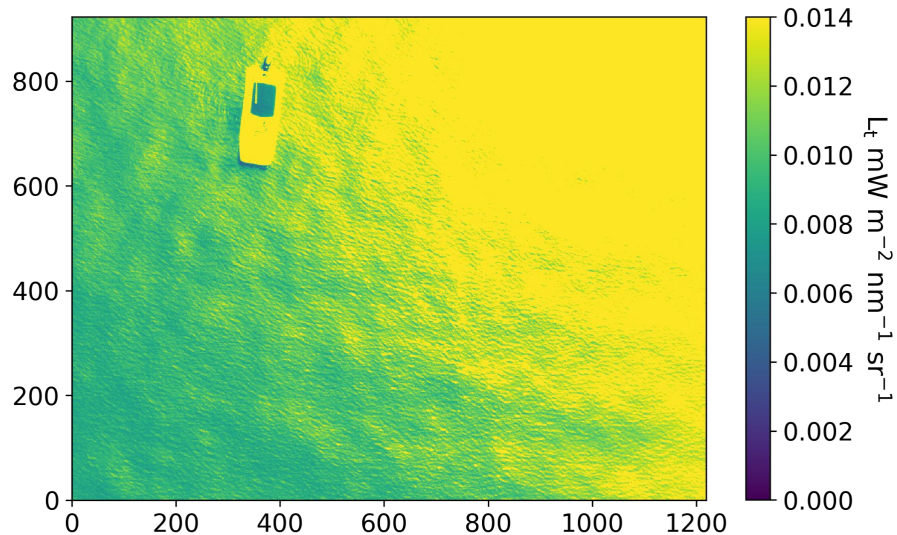




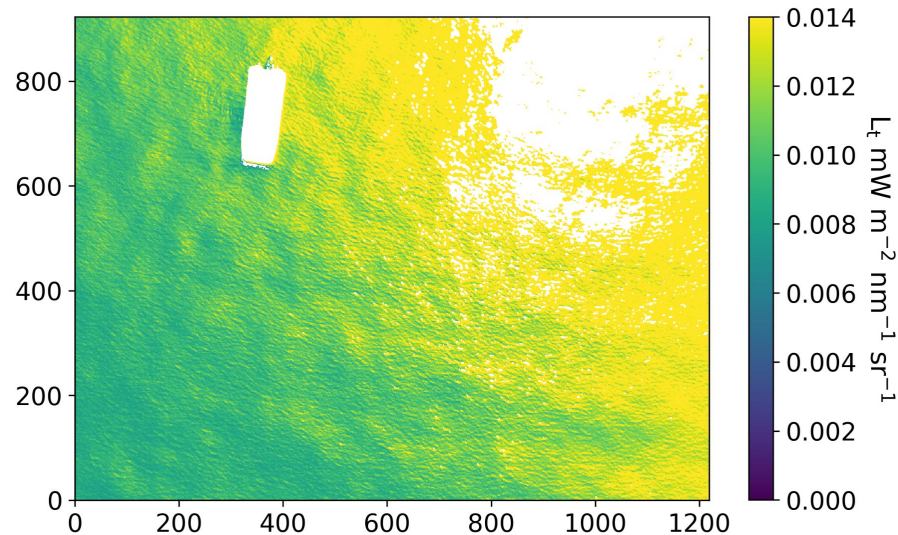
# Multispectral UAS image processing

## Step 1. Pre-processing to remove specular sun glint and non-water

Before filtering procedure



After filtering procedure



# Multispectral UAS image processing

Step 2. Apply method\* to remove surface reflected light ( $L_{SR} = L_{sky} * \rho$ )

1.  $\rho_{LUT}$

2. NIR = 0

$$R_{rs}(\theta, \Phi, \lambda) = R_{UAS}(\theta, \Phi, \lambda) - \frac{L_{sky}(\theta, \Phi, \lambda) * \rho(\theta, \Phi, \lambda)}{E_d(\lambda)}$$

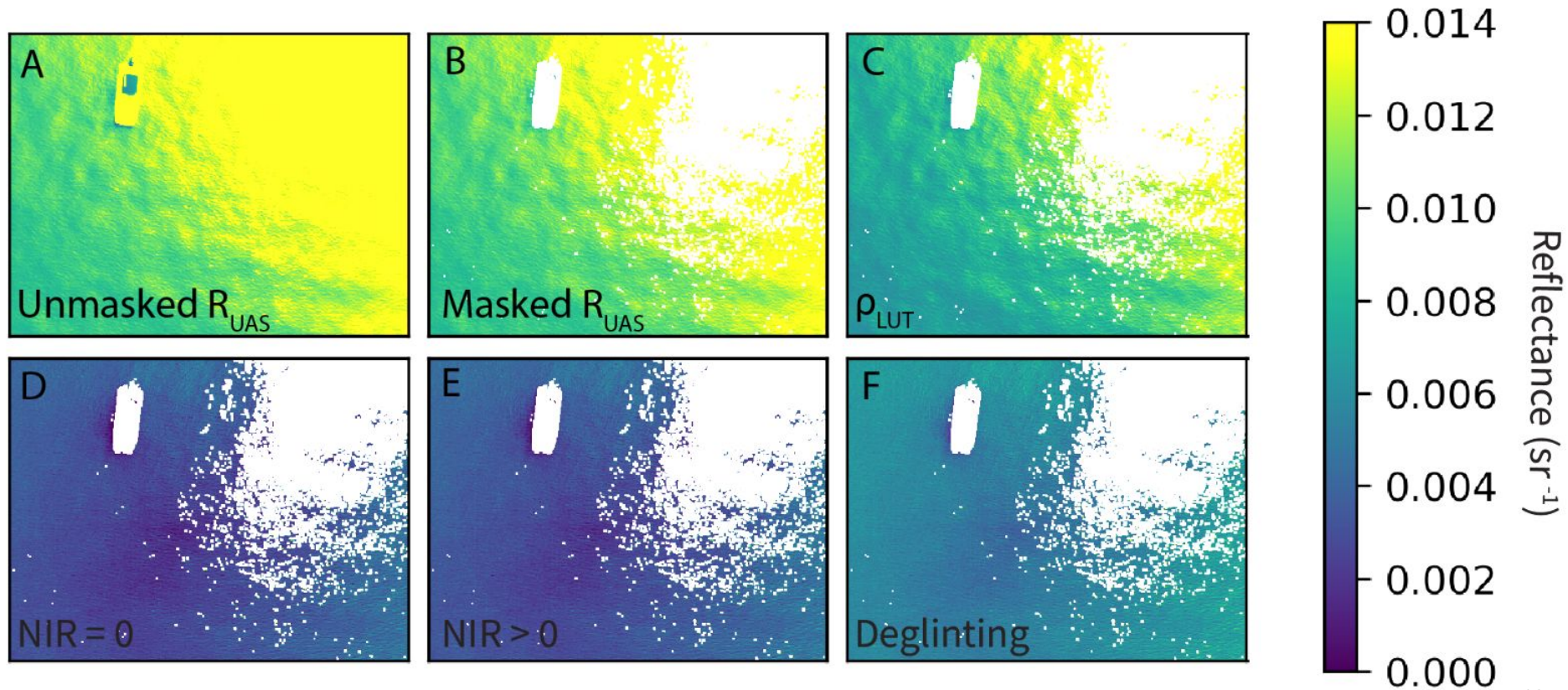
3. NIR > 0

4. Deglinting

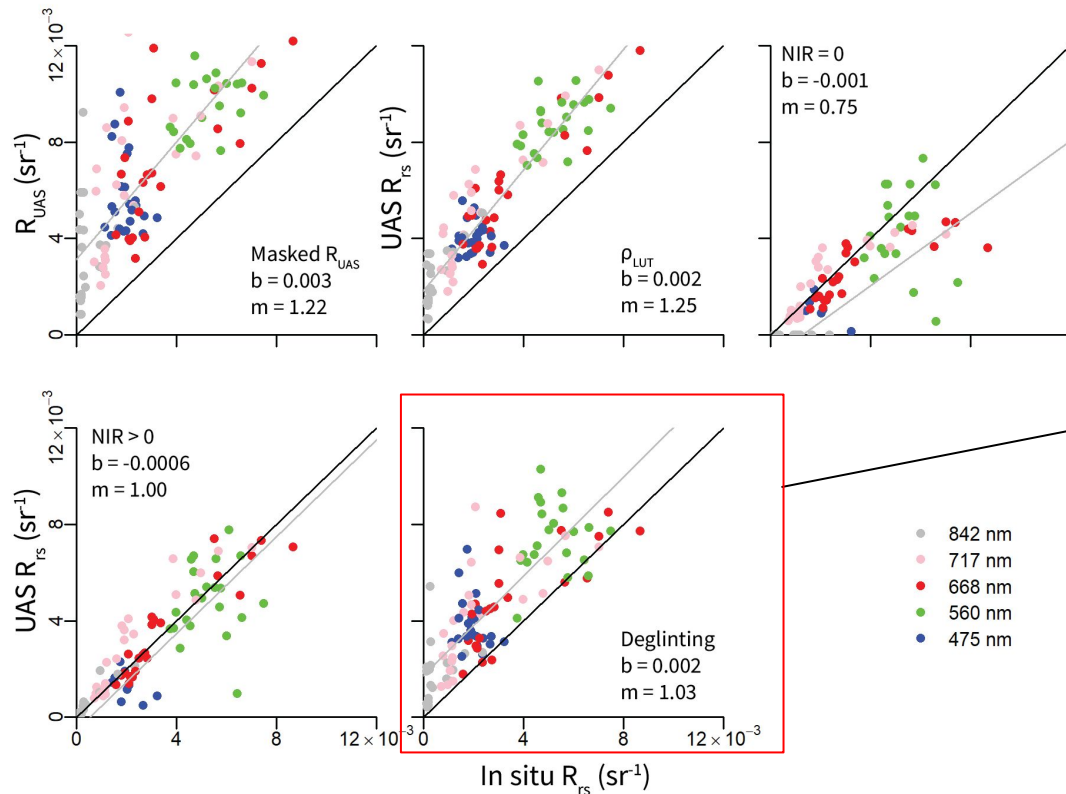
\*See Windle & Silsbe, 2021 for more details on each method



# Removing $L_{sr}$ lowers reflectance values

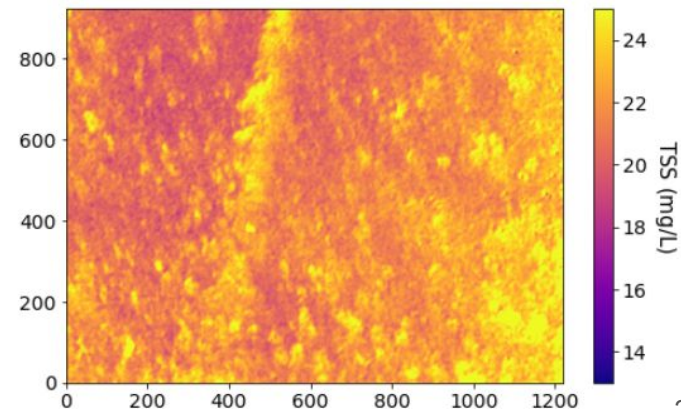
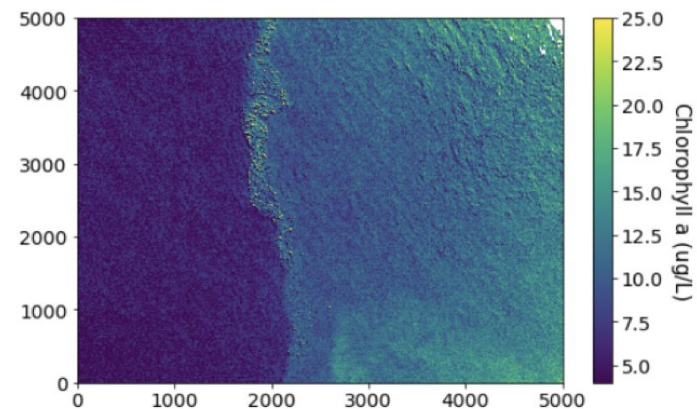
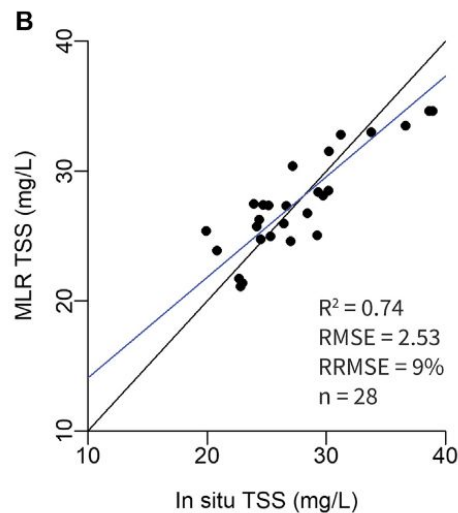
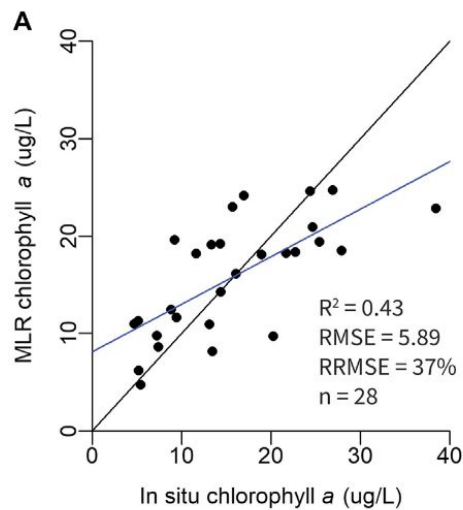


# UAS vs in situ $R_{rs}$ values



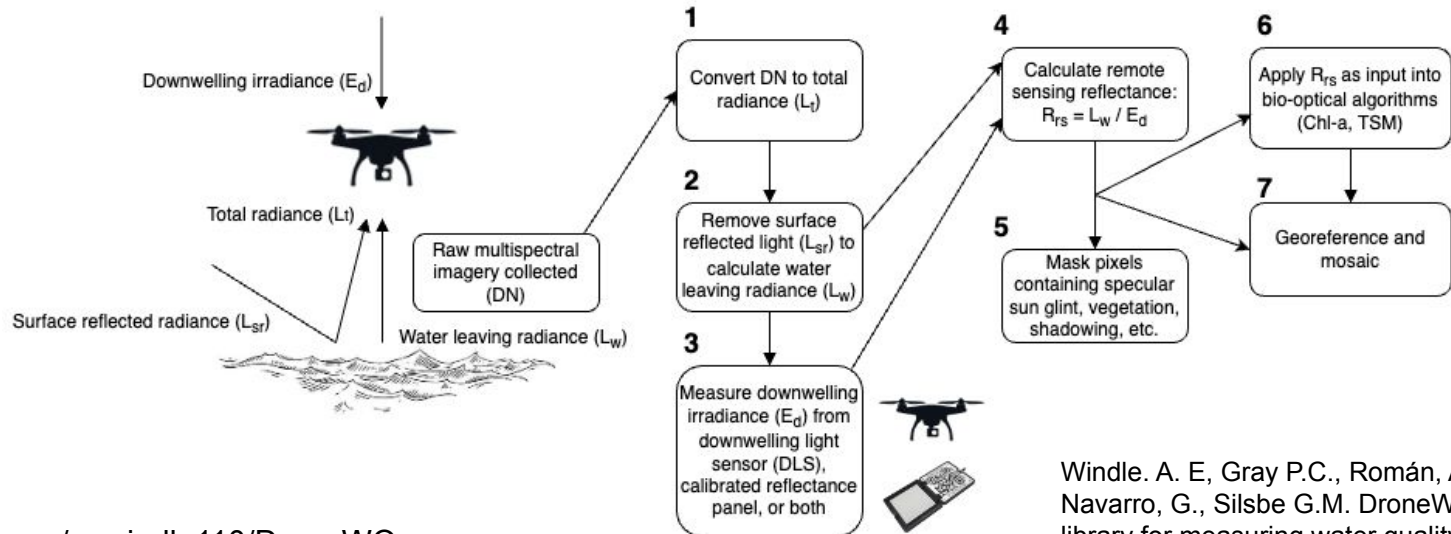
Method 4 performed best (lowest error): pixel-based approach that exploits the high absorption of water at NIR wavelengths to estimate and remove  $L_{sr}$

# High resolution water quality retrievals



# DroneWQ

Python package that can be used to analyze multispectral data collected from a drone to derive ocean color radiometry and water quality properties. Scripts are specific for the MicaSense RedEdge and Altum cameras.



<https://github.com/aewindle110/DroneWQ>  
<https://dronewq.readthedocs.io/>

Windle, A. E, Gray P.C., Román, A., Heredia, S, Navarro, G., Silsbe G.M. DroneWQ: A Python library for measuring water quality with a multispectral drone sensor. *Submitted: Journal of Open Source Software.*



## Summary

The Lake Erie drone dataset can be found in the [Zenodo DOI](#), please note it is 5.84 GB unzipped. Depending on computer speed, you may need to subset the data before running the code workflow shown here.

## Contents

1. Setup
2. View metadata
3. Process raw to Rrs
4. Convert to point samples
5. Apply bio-optical algorithms
6. Georeference and mosaic

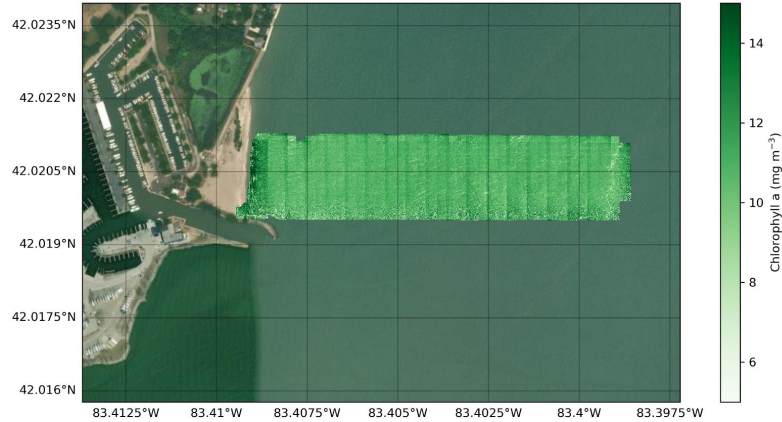
## 1. Setup

Pull in all the libraries needed for this notebook.

```
[1]: import os
```



# Georeferencing and mosaicking is challenging



Processed with DroneWQ



Article

## Enhancing Georeferencing and Mosaicking Techniques over Water Surfaces with High-Resolution Unmanned Aerial Vehicle (UAV) Imagery

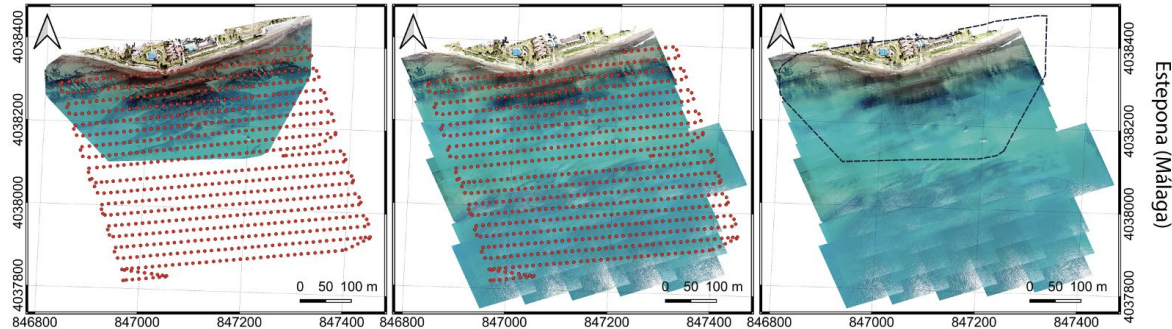
Alejandro Román <sup>1,\*</sup>, Sergio Heredia <sup>1</sup>, Anna E. Windle <sup>2,3</sup>, Antonio Tovar-Sánchez <sup>1</sup> and Gabriel Navarro <sup>1</sup>

<sup>1</sup> Department of Ecology and Coastal Management, Institute of Marine Sciences of Andalusia (ICMAN-CSIC), Spanish National Research Council (CSIC), 11510 Puerto Real, Spain; sergio.h.c@csic.es (S.H.); a.tovar@csic.es (A.T.-S.); gabriel.navarro@icman.csic.es (G.N.)

<sup>2</sup> NASA Goddard Space Flight Center, Greenbelt, MD 20771, USA; anna.windle@nasa.gov

<sup>3</sup> Science Systems and Applications Inc., Lanham, MD 20706, USA

\* Correspondence: a.roman@csic.es



• Flight plan points □ Photogrammetric software mosaicked area

# Aquatic drone pubs

- Windle & Silsbe, 2021
  - doi: [10.3389/fenvs.2021.674247](https://doi.org/10.3389/fenvs.2021.674247)
- Román et al., 2024:
  - doi: [10.3390/rs16020290](https://doi.org/10.3390/rs16020290)
- Gray et al., 2022:
  - doi: [10.1002/fee.2472](https://doi.org/10.1002/fee.2472)
- Gray et al., 2022
  - doi: [10.1002/lom3.10511](https://doi.org/10.1002/lom3.10511)

## FRONTIERS IN ECOLOGY *and the* ENVIRONMENT

Reviews | [Full Access](#)

### Drones address an observational blind spot for biological oceanography

Patrick Clifton Gray✉, Gregory D Larsen, David W Johnston

First published: 15 February 2022 |  
<https://doi.org.proxy-um.researchport.umd.edu/10.1002/fee.2472> | Citations: 3



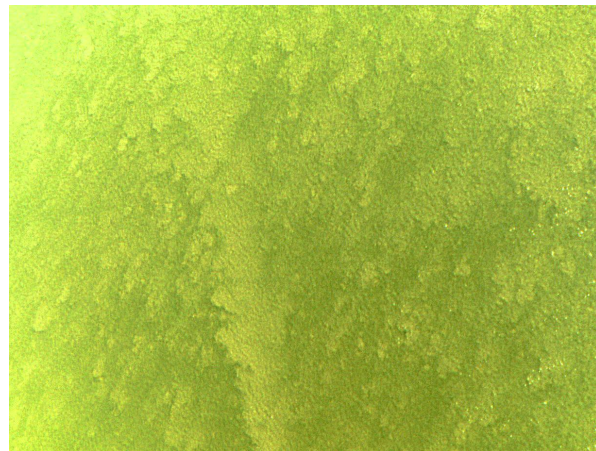
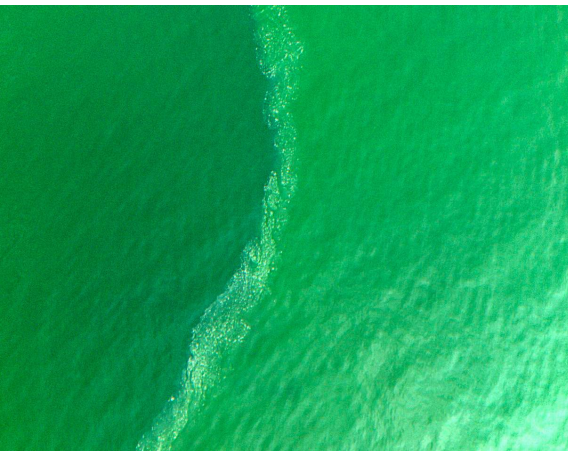
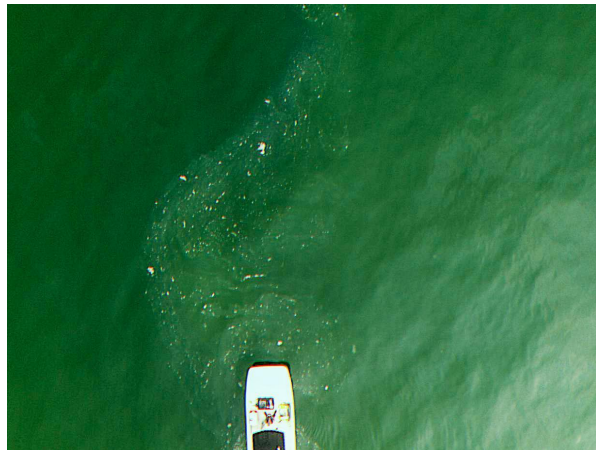
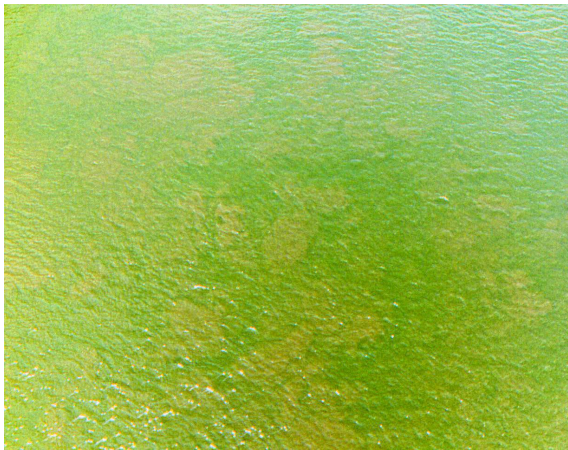
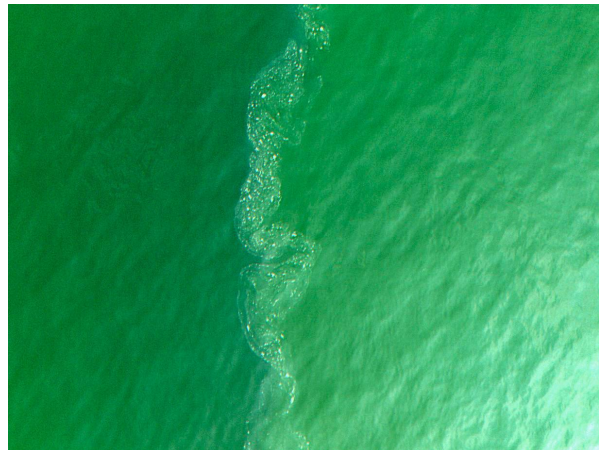
New Methods

### Robust ocean color from drones: Viewing geometry, sky reflection removal, uncertainty analysis, and a survey of the Gulf Stream front

Patrick Clifton Gray✉, Anna E. Windle, Julian Dale, Ivan B. Savelyev, Zackary I. Johnson, Greg M. Silsbe, Gregory D. Larsen, David W. Johnston



# Fine-scale observations from UAS imagery



# Future directions

- Collect hyperspectral UAS data
  - Point spectrometer vs push-broom/line scan camera
- Apply semi-analytical methods to derive phytoplankton community composition metrics
  - Assist with harmful algal bloom identification and monitoring
- Using UAS  $R_{rs}$  to validate ocean color satellites



Thanks for listening! Questions?

[anna.windledipaola@nasa.gov](mailto:anna.windledipaola@nasa.gov)

