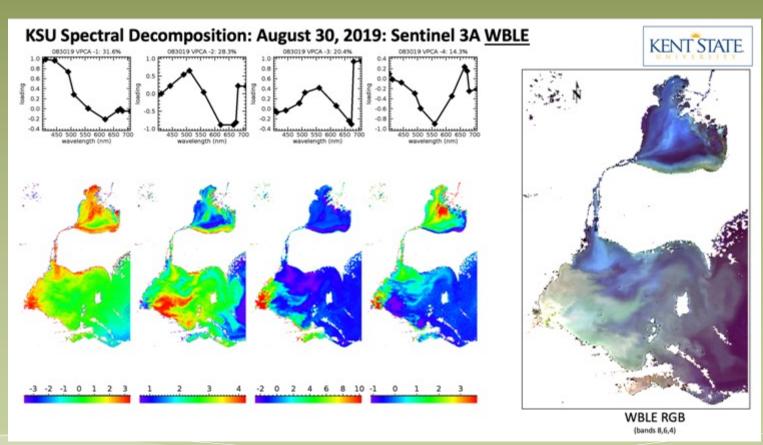
KENT STATE.

Porting VPCA to Google Earth Engine: Harmful algal blooms and Benthic cover

Dr. Joseph Ortiz (jortiz@kent.edu) Kent State University Department of Geology

Today's talk

- 1. Lake Erie
- 2. Indian River Lagoon
- 3. Examples in GEE
 - 1. Landsat 8 HABS
 - 2. Sentinel 2 HABS
 - 3. Sentinel 2 Benthic cover



KSU VPCA spectral decomposition from Sentinel 3A OLCI Lake Erie August 30, 2019

J. Ortiz, (jortiz@kent.edu)

The Kent State Approach



- Goal: Quantify the relationship between phytoplankton pigments and phytoplankton assemblages from Field Samples, Field Spectroradiometers, Remote Sensing data
- Objectives
 - Measure water samples by visible derivative spectroscopy
 - Match spectral pigment assemblages to known signatures for classes of phytoplankton
 Compare pigment assemblages to measures of concentration in the lake (Cell counts, pigments, degradation products, genomics).



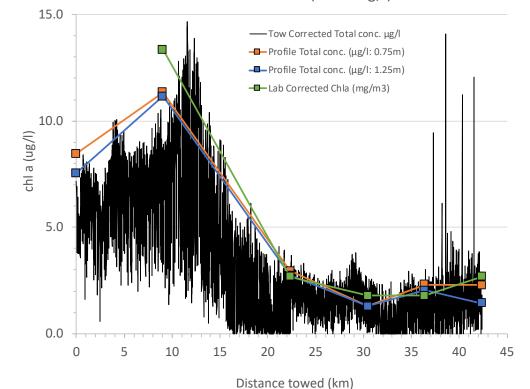
J. Ortiz, (2019 NOAA Coast Watch meeting ; jortiz@kent.edu)

Lake Erie 9/26/18, Detroit Plume Transect Sampling



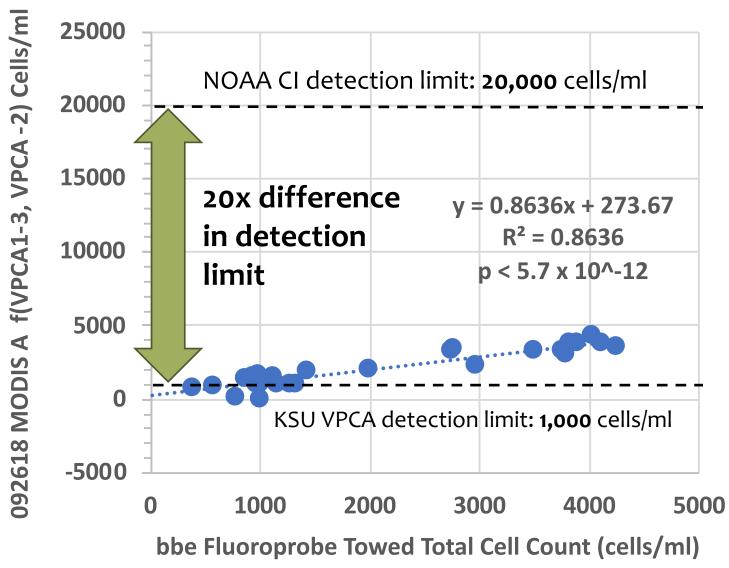


092618 DPT Total concentration (chl a : ug/l)

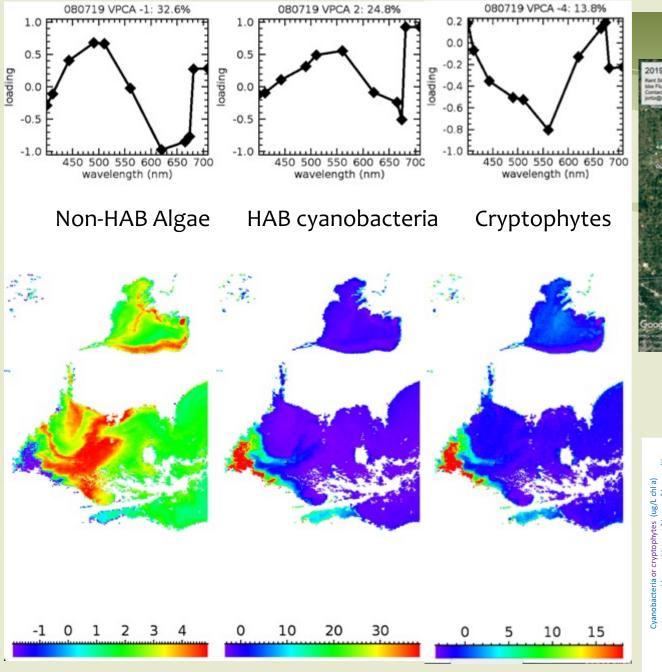


J. Ortiz, (2019 NOAA Coast Watch meeting ; jortiz@kent.edu)

092618 bbe Fluoroprobe Tow Corrected Total cell count (1k running avg) Actual vs. Estimates







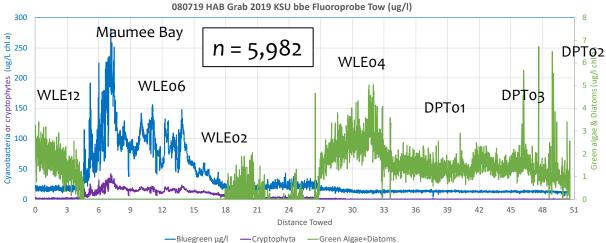




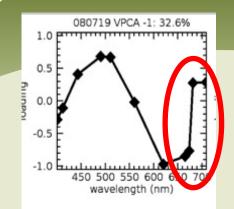
2019 HAB Grab Kent State University Dr. Joseph D. Ortiz jortiz@kent.edu



* Approx. sample locations

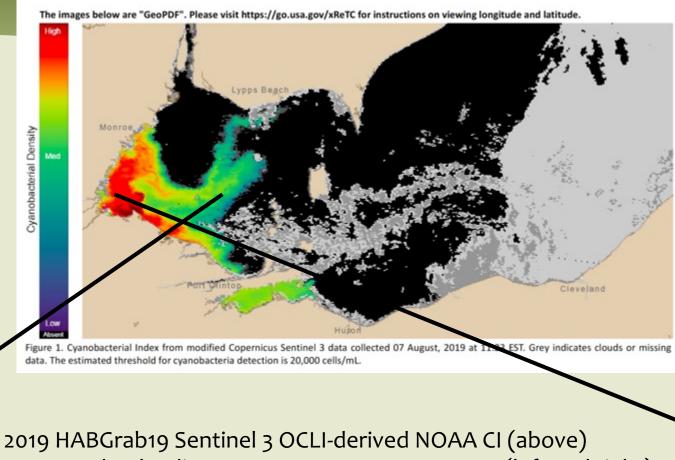


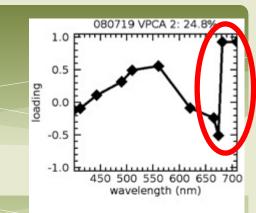
Standardized units



-1 0

2 3





2019 HABGrab19 Sentinel 3 OCLI-derived NOAA CI (above) compared to leading two KSU VPCA components (left and right) The NOAA CI represents a mixture of these two components and potentially two additional components (not shown) for NOAA Threshold of 20K cells/ml.



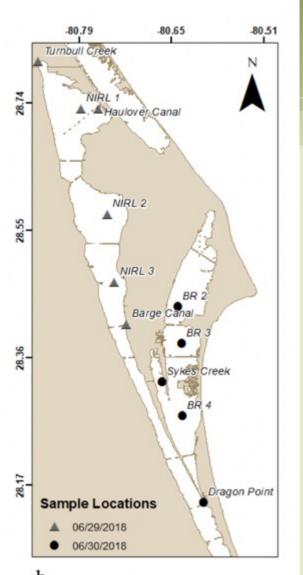


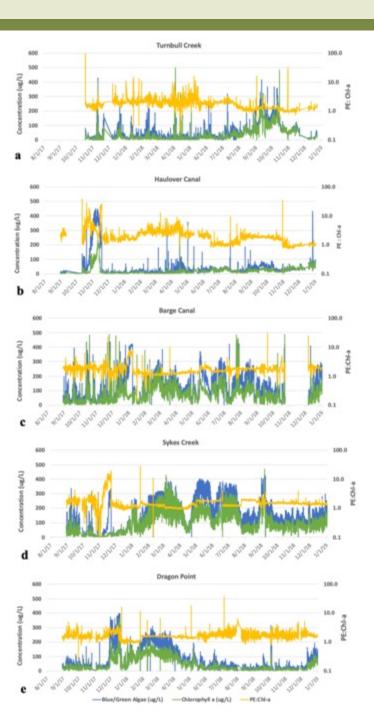
J. Ortiz, (2019 NASA Glenn HAB Algorithm meeting ; jortiz@kent.edu)

Application to IRL with ORCA





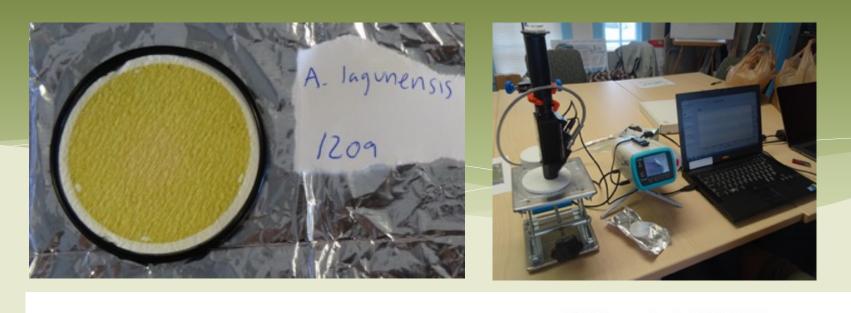


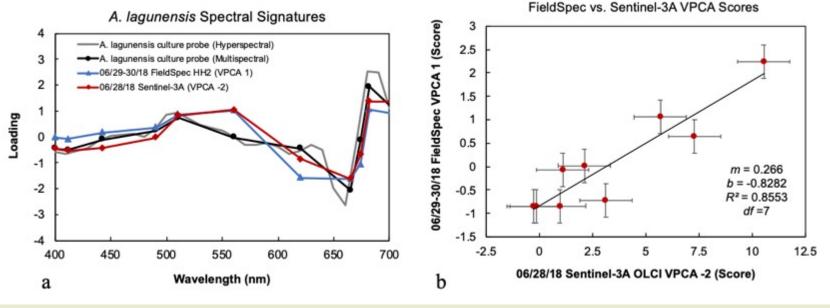


A. Lagunensis culture ID

- Spectral fingerprint from an A. lagunensis culture matches field and satellite spectra
- Relative abundance from field and satellite station data match



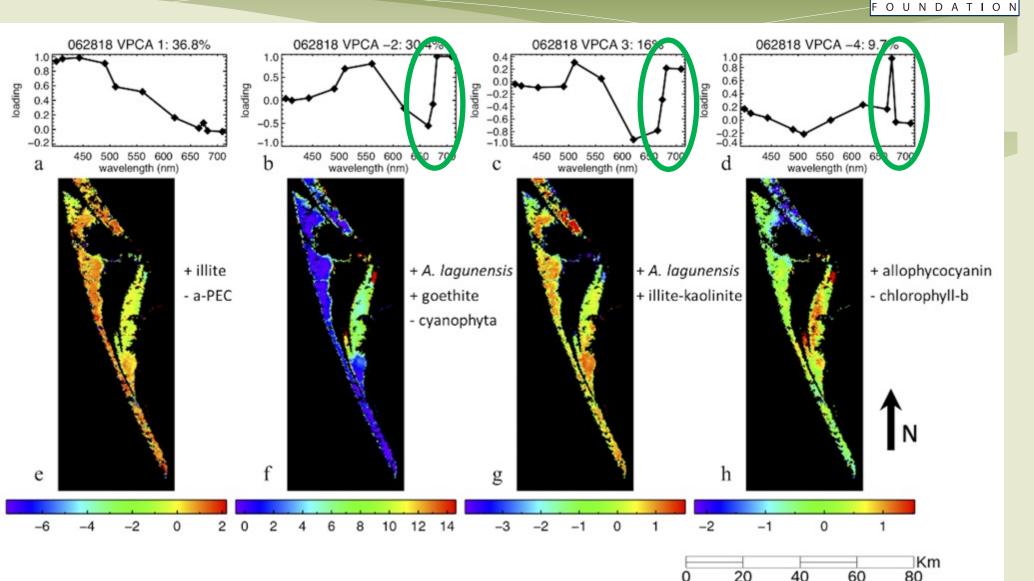


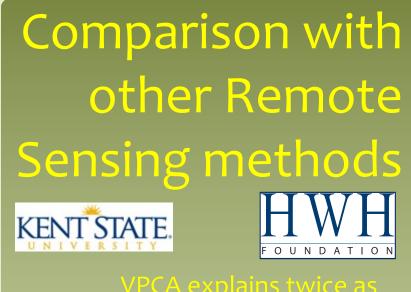


KENT STATE Chlorophyll a prediction in the Indian River Lagoon ${ m HWH}$

 KSU method splits the image variance

Select subset
 of components
 with a pigment
 related Red edge response





VPCA explains twice as much variance with one third the noise

Published NCDI Calibration

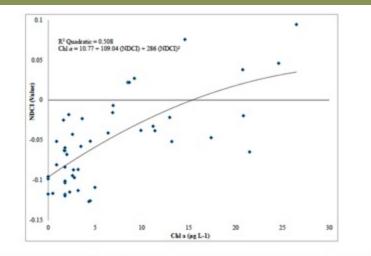
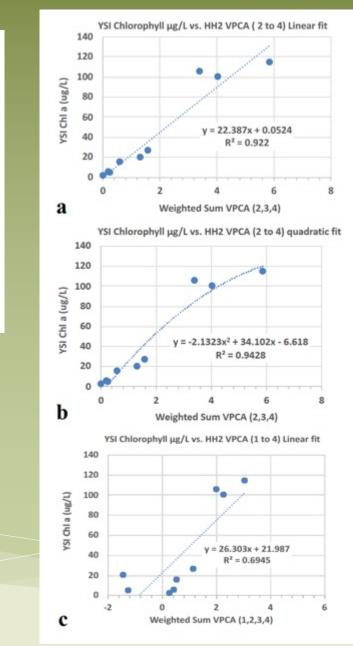


Figure 4. Regression analysis of *in-situ* Chl *a* measurements and their corresponding MERIS pixel data that has been converted by the NDCI algorithm. The resulting line of best fit is used to convert the NDCI values into estimated Chl *a*.

Table 5. Comparison to Kamerosky et al., Remote Sensing, vol. 7, 1441-1460, 2015 Calibration results

| | | NDCI Quadratic | This Study (Weighted S3A | This Study (Weighted S3A | This Study (Weighted S3A |
|-----------------|------------------|--------------------|----------------------------------|------------------------------------|----------------------------------|
| | Moses et al., | Mishra & Mishra | OLCI VPCA 2 to 4, Linear fit) | OLCI VPCA 2 to 4,Quadratic fit) | OLCI VPCA 1 to 4, Linear fit) |
| Statistic | (2009) | (2011) | Model 1 | Model 2 | Model 3 |
| R | 0.66 | 0.71 | 0.96 | 0.97 | 0.83 |
| \mathbb{R}^2 | 0.43 | 0.51 | 0.92 | 0.94 | 0.69 |
| RMSE (ug/L) | 37.42 | 34.93 | 13.38 | 11.47 | 39.9 |
| Std. Err (ug/L) | 5.29 | 4.94 | 4.46 | 3.82 | 13.31 |
| n | 50 | 50 | 9 | 9 | 9 |

VPCA Calibration



Comparison of individual daily VPCA Results with average VPCA results.



14

062818 S3A VPCA -2

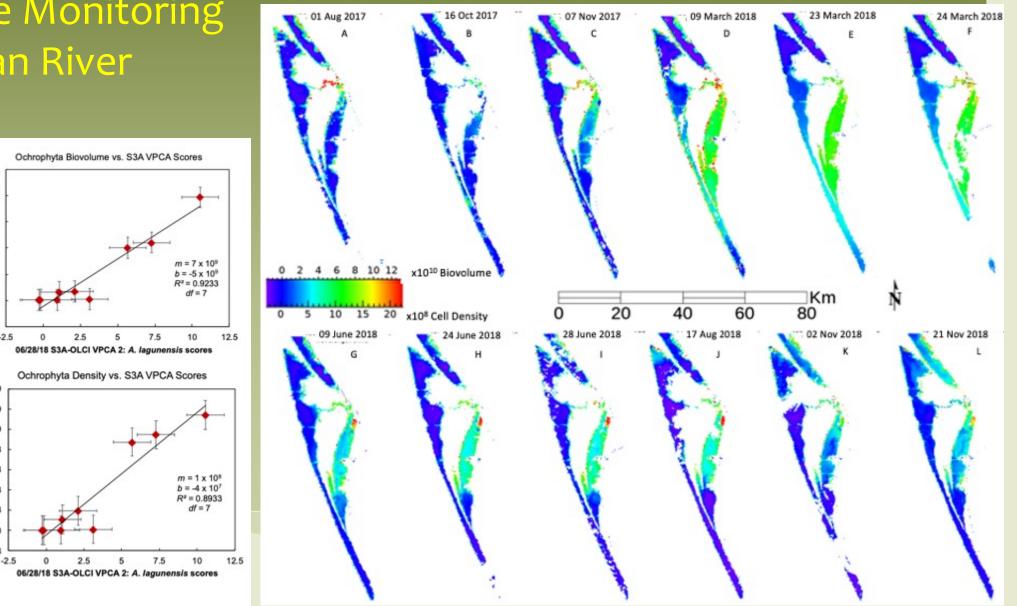
IRL S3A-OLCI VPCA Pattern B: A. lagunensis scores

- The extracted daily component loadings are stable, so the component scores and the regression coefficients of the average component loadings are stable.
- Averaging the component

 loadings gives even cleaner
 transfer functions to cell
 density and biovolume than the
 daily estimates: R² increases to
 0.96 and 0.94

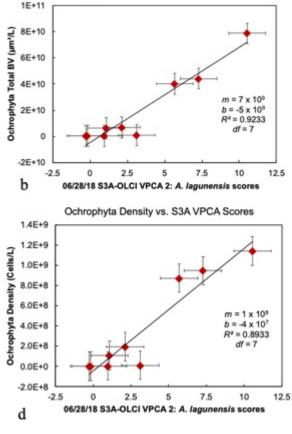
Pattern B Spatial Signature vs. S3A IRL Pattern B VPCA Scores 14 12 s3A VPCA -2 * * 01 062818 v = 0.9124x + 0.2012 $R^2 = 0.988^2$ 12 10 **IRL S3 VPCA Pattern B** Ochrophyta Density vs. S3A VPCA Scores Ochrophyta Biovolume vs. S3A VPCA Scores 1.4E+9 1E+11 1.2E+9 (http://www.setator 6E+10 6E+10 (Cells/L) 1.0E+9 8.0E+8 Total I 4E+10 6.0E+8 Ochrophyta ohvta 4.0E+8 2E+10 2.0E+8 2803E+09x - 3.7802E+09 v = 1.1354E+08x - 3.0483E+07 R² = 9 3529E-01 0.0E+0 -2E+10 -2.0E+8 -2 12 14 -2 12

IRL S3A-OLCI VPCA Pattern B: A. lagunensis scores



Brown Tide Monitoring in the Indian River Lagoon

FOUNDATION KENT STATE.



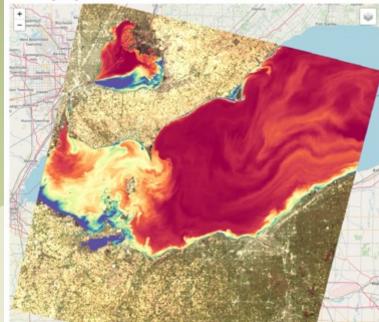
Lake Erie LC08_019031_20200420: RGB



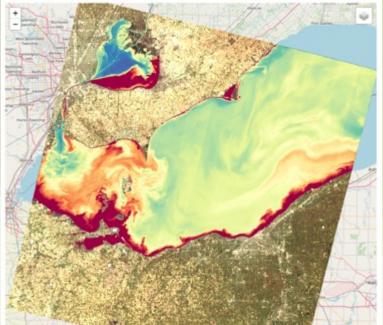
VPCA in Google Earth Engine

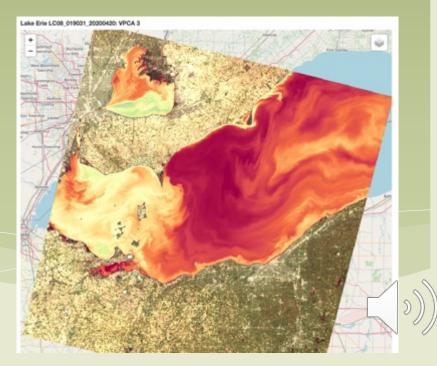
Lake Erie Landsat 8 WRS-2 Scene 019 031 on 20200420

Lake Erie LC08_019031_20200420: VPCA 1



Lake Erie LC08_019031_20200420: VPCA 2





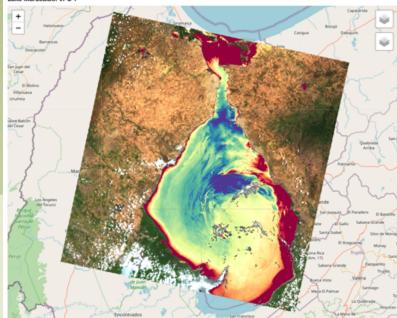




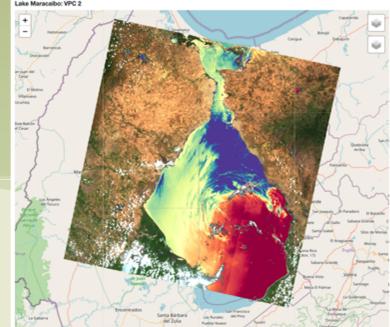
VPCA in Google Earth Engine

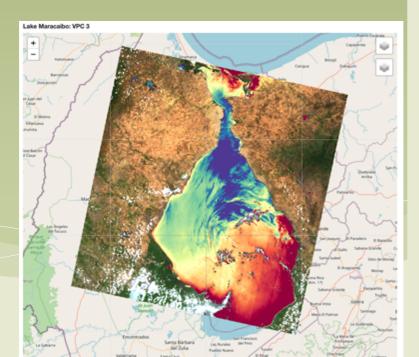
Lake Maracaibo Landsat 8 WRS-2 Scene 007 053 on 20200416

ake Maracaibo:



Lake Maracalbo: VPC 2







DG DEFIS #StrongerTogether ② @defis_eu · Jan 16 #ImageOfTheDay

Lake Valencia is the largest freshwater lake in #Venezuela

Its **#WaterQuality** is affected by the influx of untreated wastewater from the surrounding urban, agricultural & industrial land uses

...

Massive algal bloom seen by @Copernicus #Sentinel2



Lake Valencia RGB 2021-01-12



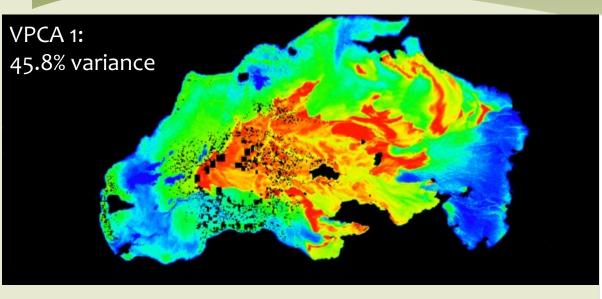
Recurrent cyanobacterial blooms on Lake Valencia

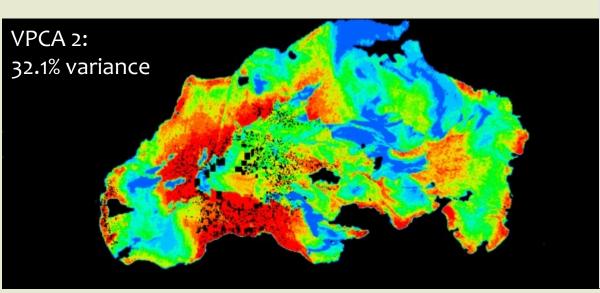
Lake Valencia RGB 2021-12-15

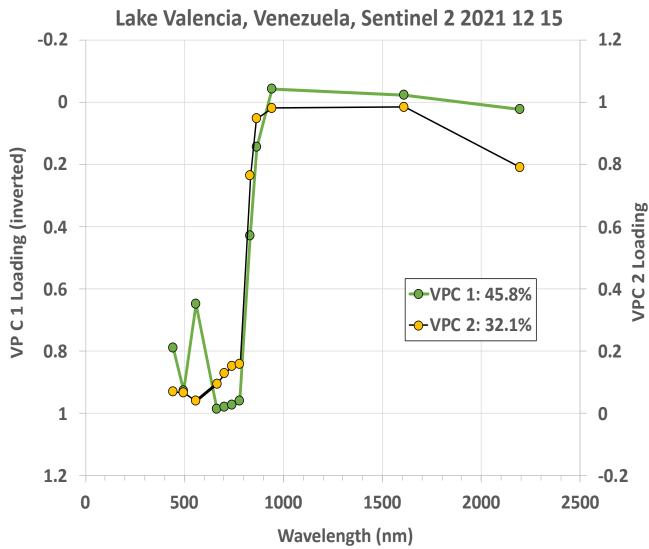
Lake Valencia also has a Complex algal community including green algae, diatoms



KSU Spectral Decomposition: Leading two components: 78% image variance

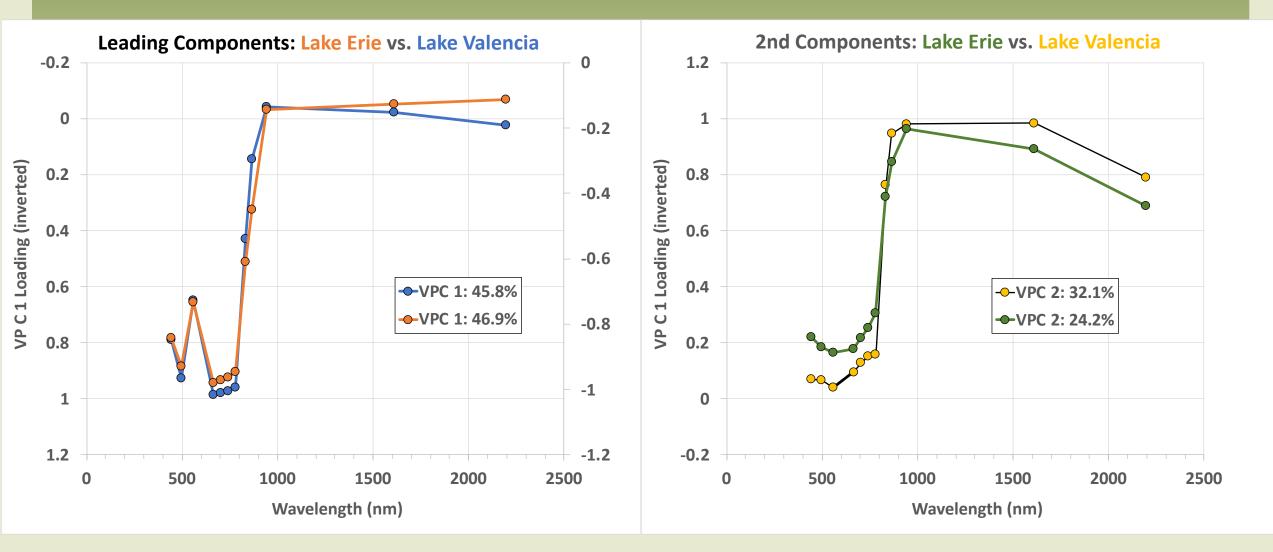




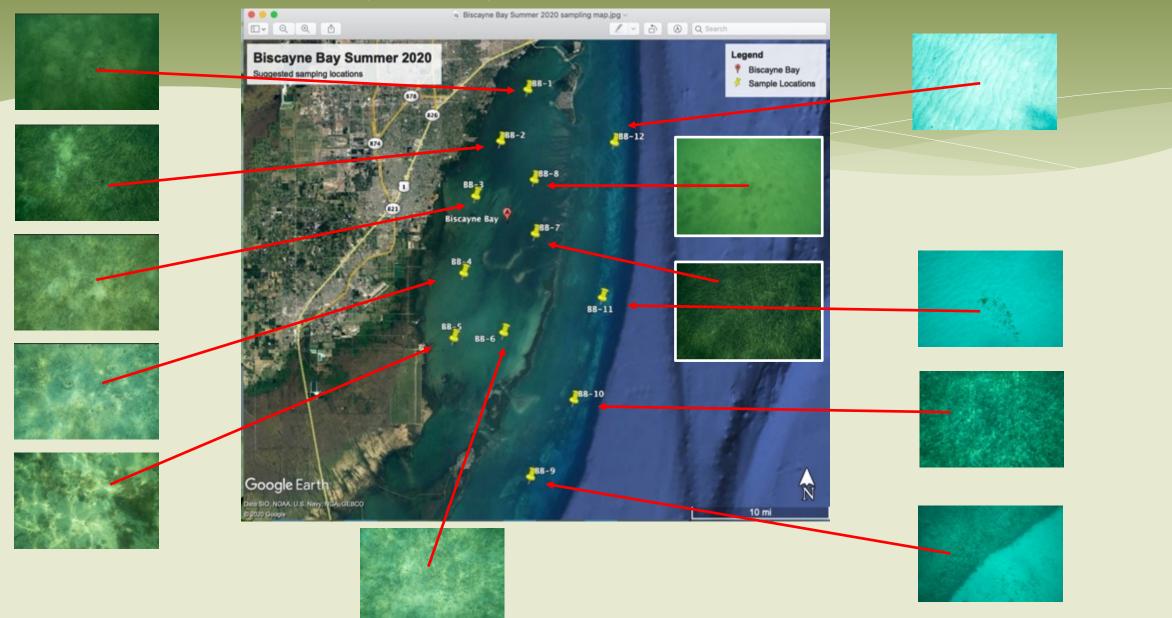


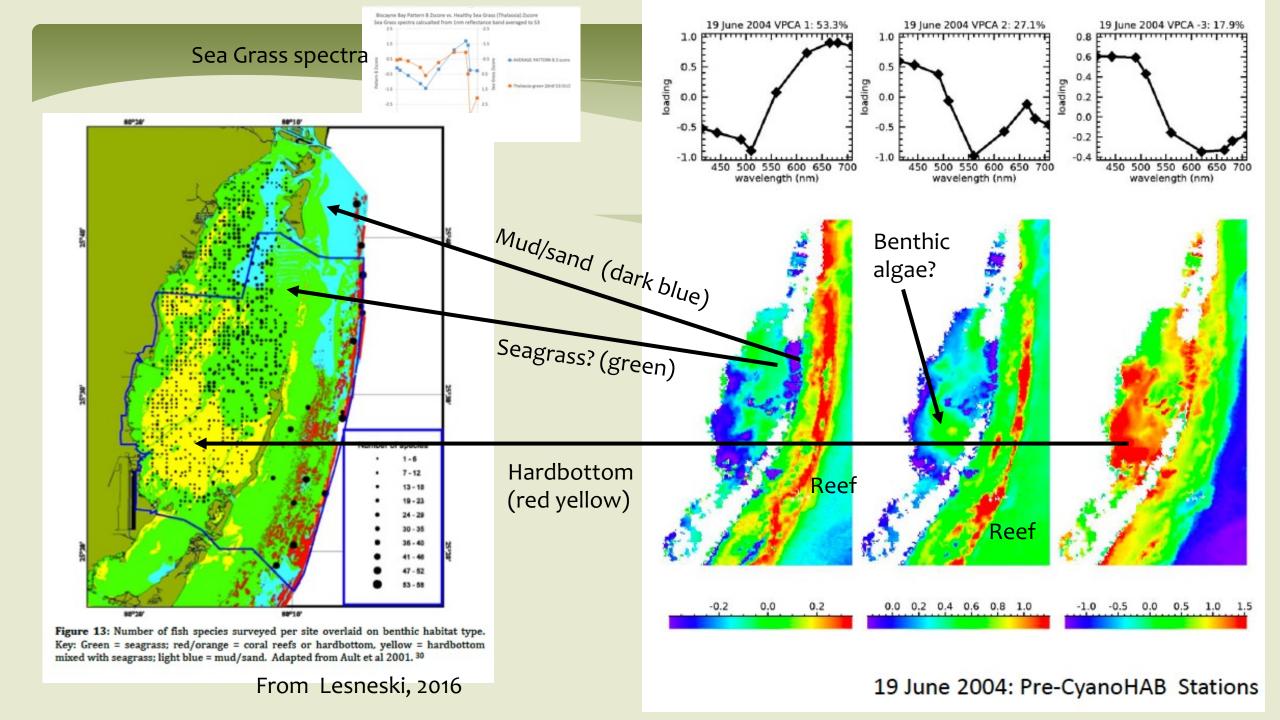
jortiz@kent.edu

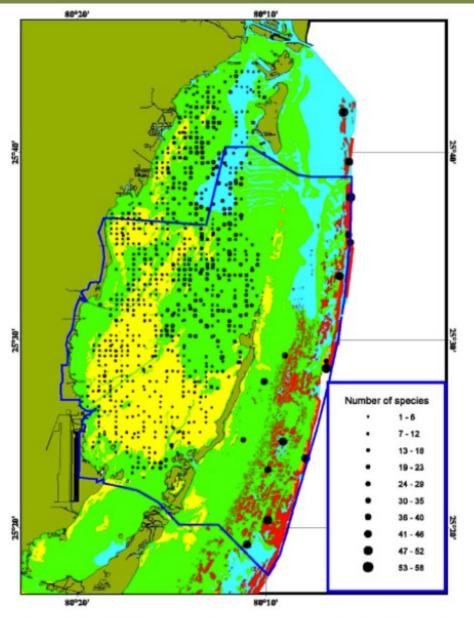
Matching components



Fall 2020 Biscayne Bay Benthic Cover Nadir views







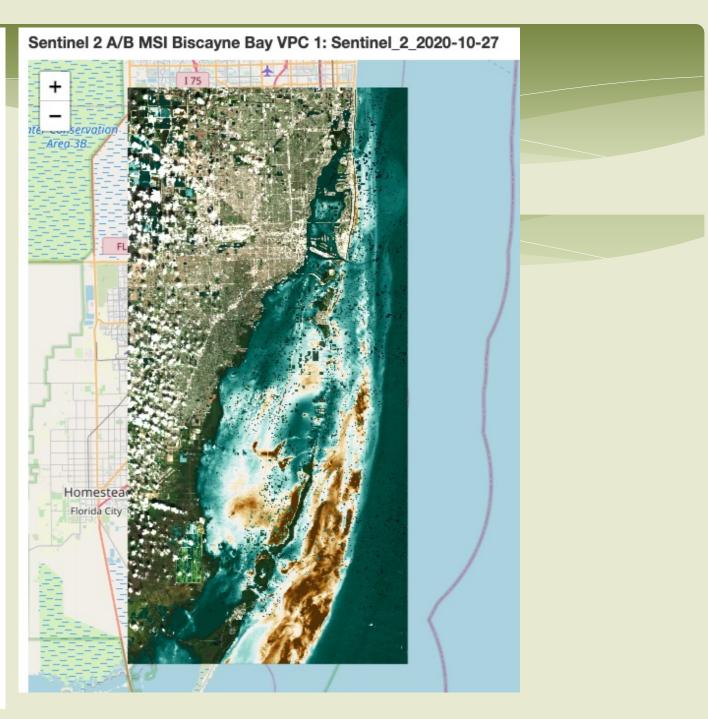


Figure 13: Number of fish species surveyed per site overlaid on benthic habitat type. Key: Green = seagrass; red/orange = coral reefs or hardbottom, yellow = hardbottom mixed with seagrass; light blue = mud/sand. Adapted from Ault et al 2001.³⁰



Summary and implications



- Method can provide a variety of products using NASA and ESA sensors: Phytoplankon classes, pigment degradation products, suspended minerals
- VNIR derivative spectroscopy unmixes and quantifies plant pigment assemblages in Optically complex aquatic systems (e.g. Lake Erie, Indian River Lagoon)
- Tracks phytoplankton contribution to eutrophication, with implications for harmful algal blooms, anoxia, and fisheries
- KSU VPCA decomposition method has be applied successfully to hyperspectral and multispectral lab samples, field-based spectroradiometers, HICO, NASA Glenn HSI2, MODIS A/T, Landsat 4-8, Sentinel-3A/B
- VPCA is in the process of being ported to GEE: Preliminary results are very promising
- VPCA is well suited for application to current and the upcoming hyperspectral SBG and PACE missions: Makes use of <u>all</u> information present in hyperspectral data

Ortiz et al., (2015 HAB Data meeting; jortiz@kent.edu)

Recent Publications

- See Water quality webpage at: http://www.personal.kent.edu/~jortiz/home/wqr.html
- Judice, T., Widder, E.A., Falls, W.H., Avouris, D.M., Cristiano, D.J., & Ortiz, J.D. Field-validated detection of Aureoumbra lagunensis brown tide blooms in the Indian River Lagoon, Florida using Sentinel-3A OLCI and ground-based hyperspectral spectroradiometers, GeoHealth, 2020. https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/2019GH000238
- Ortiz, J.D., Avouris, D. Schiller, S.J., Luvall, J.C., Lekki, J.D., Tokars, R.P., Anderson, R.C., Shuchman, R., Sayers, M. and R. Becker, (2019) Evaluating visible derivative spectroscopy by varimax-rotated, principal component analysis of aerial hyperspectral images from the western basin of Lake Erie, Journal of Great Lakes Research, Volume 45, Issue 3, June 2019, Pages 522-535, https://doi.org/10.1016/j.jglr.2019.03.005
- Avouris, D.M. and Ortiz, J.D., (2019) Validation of 2015 Lake Erie MODIS image spectral decomposition using visible derivative spectroscopy and field campaign data, Journal of Great Lakes Research Volume 45, Issue 3, June 2019, Pages 466-479, https://doi.org/10.1016/j.jglr.2019.02.005
- Ortiz, J.D., D. Avouris, S. Schiller, J. Luvall, J. Lekki, R.P. Tokars, R.C. Anderson, R. Shuchman, M. Sayers, and R. Becker, Intercomparison of Approaches to the Empirical Line Method for Vicarious Hyperspectral Reflectance Calibration, Front. Mar. Sci., vol. 4, 14 September 2017, <u>https://doi.org/10.3389/fmars.2017.00296</u>
- Ali, K.A., and J.D. Ortiz, Multivariate approach for chlorophyll-a and suspended matter retrievals in Case II waters using hyperspectral data, Hydrological Sciences Journal, 2014. DOI 10.1080/02626667.2014.964242.
- Ortiz, J.D., Witter, D.L., Ali, K.A., Fela, N., Duff, M., and Mills, L., Evaluating multiple color producing agents in Case II waters from Lake Erie, International Journal of Remote Sensing, 34 (24), 8854-8880, 2013.
- Mou, X, Jacob, J., Lu, X., Robbins, S., Sun S., J.D. Ortiz. Diversity and distribution of free-living and particle associated bacterioplankton in Sandusky Bay and adjacent waters of Lake Erie Western Basin, Journal of Great Lakes Research 2013.
- Ali, K.A., Witter, D.L., and J.D. Ortiz, Application of empirical and semi-analytical algorithms to MERIS data for estimating chlorophyll a in Case waters of Lake Erie, Environmental Earth Sciences; DOI 10.1007/s12665-013-2814-0, published Oct 1, 2013.
- Ali, K.A., Witter, D.L., and J.D. Ortiz, 2012, Multivariate approach to estimate color producing agents in Case 2 waters using first-derivative spectrophotometer data, Geocarto International, 10/30/2012 DOI:10.1080/10106049.2012.743601.
- Witter, D., Ortiz, J.D., Palm, S. Heath, R., Budd, J., Assessing the Application of SeaWiFS Ocean Color Algorithms to Lake Erie, Journal of Great Lakes Research, 35, 361-370, 2009.