

BioREaCM: Biodiversity-Remote sensing for Estuarine and Coastal Habitat research

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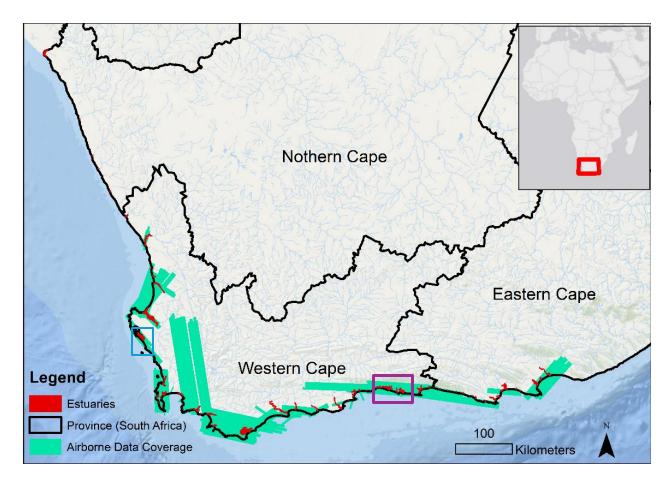






Project Background: BioR = 3 CH

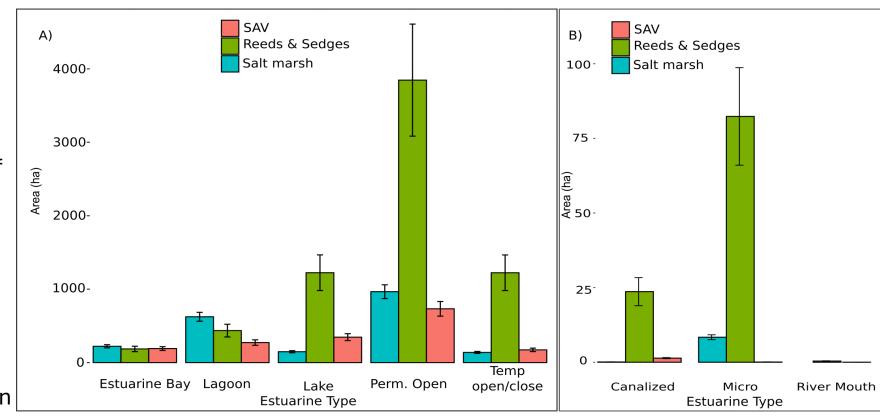
- Biodiversity-Remote sensing for Estuarine and Coastal Habitat research(BioREaCH)is focused on understanding the biodiversity of coastal wetlands across the Greater Cape Floristic Region of South Africa using imaging spectroscopy, LiDAR, and multispectral satellite data.
- Recent efforts by collaborators to map change across coastal wetlands in South Africa found that medium resolution satellites overpredicted both seagrass and salt marsh habitats (Van Deventer et al. 2025)



Satellite data baseline

- We mapped 10,848.9 ha ± 1830.9 of coastal wetlands (16% of total extent of the estuaries)
 - Reeds and Sedges: 7,022.7 ±1,392.8 ha (64.7%)
 - Salt marsh: 2,107.6 ± 203.9 ha (19.4%)
 - SAV: 1,718.6 ± 234.25 ha (15.8%)

We mapped 9,066.35 ±2,154.6 ha of tidal marsh slightly higher than the global tidal marsh layer's estimate of 8,026.34 ha for the same 83 estuaries.



Campbell et al. in revision

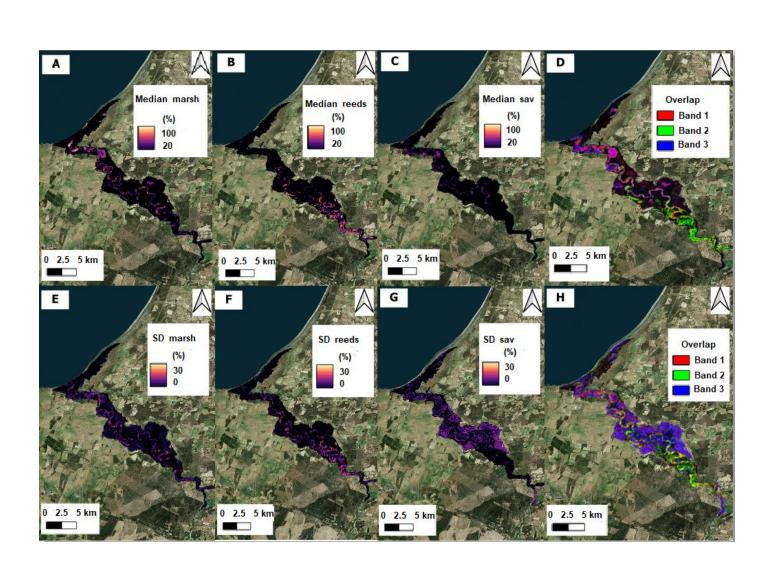


EMIT

Assessed the NASA Earth Surface Mineral Dust Source Investigation (EMIT) sensor for utility in mapping coastal wetland extent.

- Evaluated three wetland habitats at the sub-pixel scale
- Best performance was for low tidal stage images during the growing season.

Thakali, Campbell and Adam, in Revision

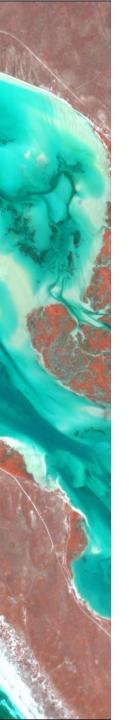




Fieldwork

- 64 10 m vegetation plots
 - Identified species
 - % cover
 - Vegetation height
- Species and plot level spectral data
- 19 soil cores
- Collected within 2 weeks of airborne acquisitions
- >500 GPS vegetation community locations





BioR E a C H updates

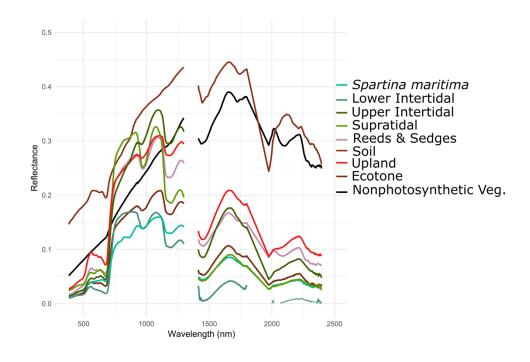
- Spectral library development and application for Plant Functional Type mapping
- 2. Estuarine EBVs salinity
- 3. Hydrological modeling



Spectral libraries

- We used three hundred and ten *in situ* ASD FieldSpec measurements to create our spectral library
- Develop a spectral library and apply to airborne data for plant functional community classification

- Broad
 Water, Wetland, Upland, Soil
- Tidal marsh community
 Water, Supratidal marsh, Upper intertidal marsh, Intertidal marsh, Upland, Reeds and Sedges, Soil





MESMA

Multiple Endmember Spectral Mixture Analysis (MESMA; Ochoa et al. 2025) was used to apply the spectral signatures to AVIRIS-NG data collected during the BioSCape campaign.

MESMA assumes each pixel is composed of a combination of pure spectra or endmembers.

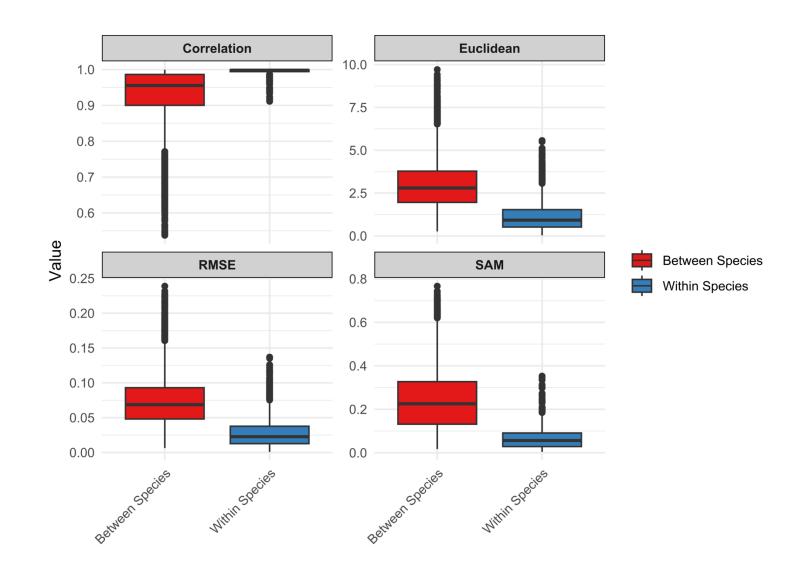
In our study, the spectral library is composed of pure spectra collected in the field in Langebaan Lagoon coincident with airborne image acquisitions.



Evaluating in situ spectra

Compare metrics of similarity to identify spectra that were inconsistent with those from the same species.

Metrics: Euclidean distance, RMSE, Spectral Angle Mapping, and correlation

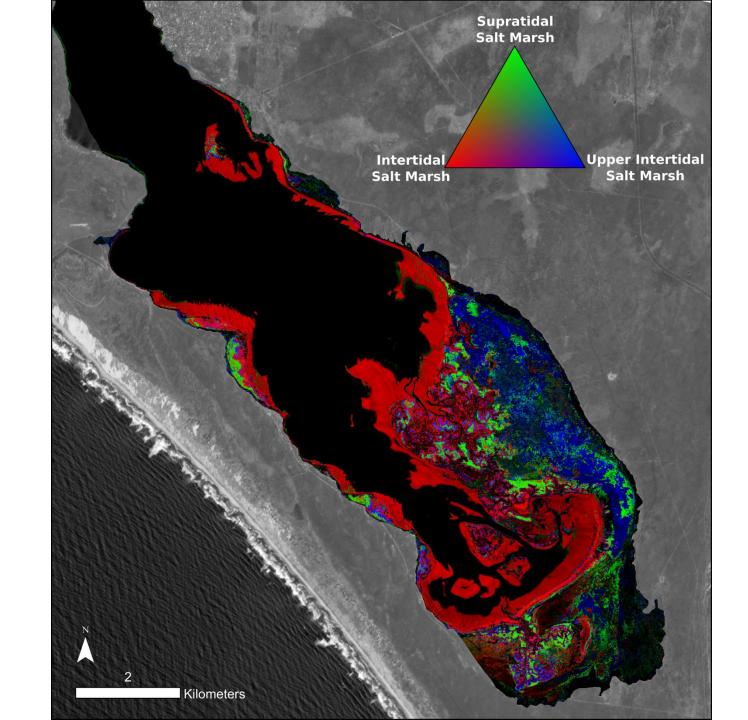




Langebaan Lagoon – 10/29/2023 (0.077 m above MSL)

Salt marsh habitats

 Intertidal is over predicted

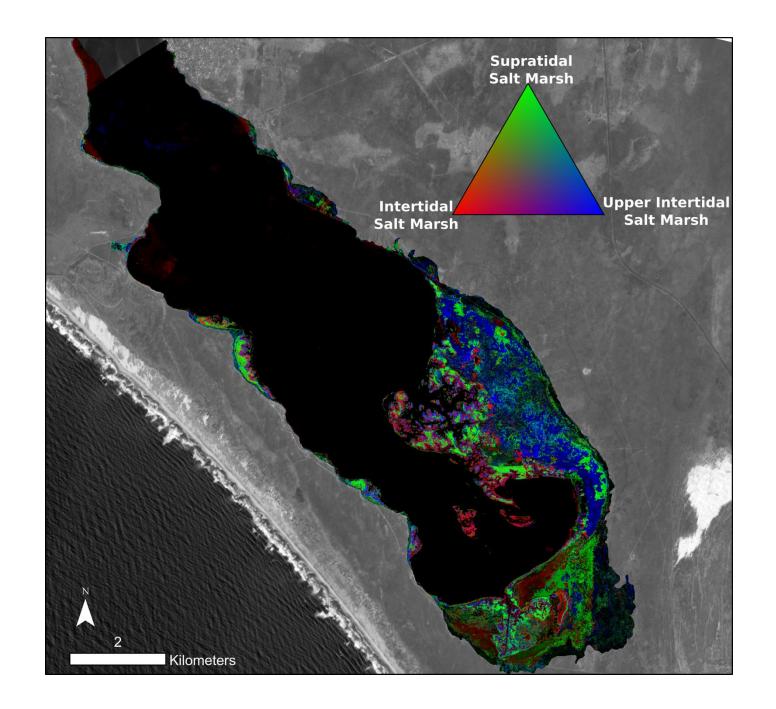




Langebaan Lagoon – 10/26/2023 (0.708 m above MSL)

Salt marsh habitats

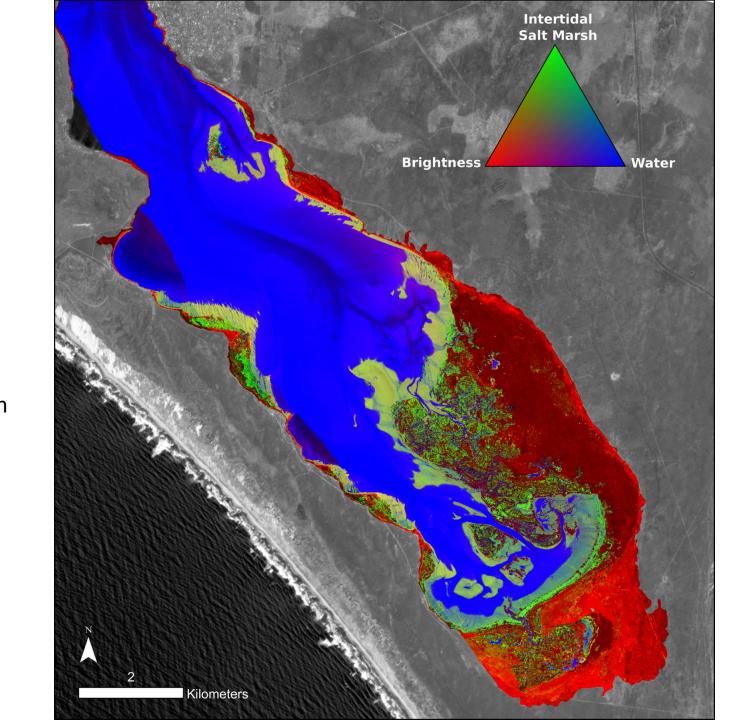
 Intertidal extent reduced due to higher water levels and no submerged spectra in the spectral library

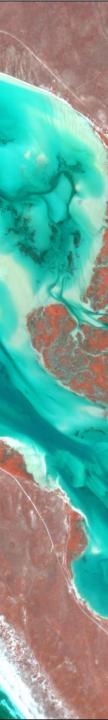




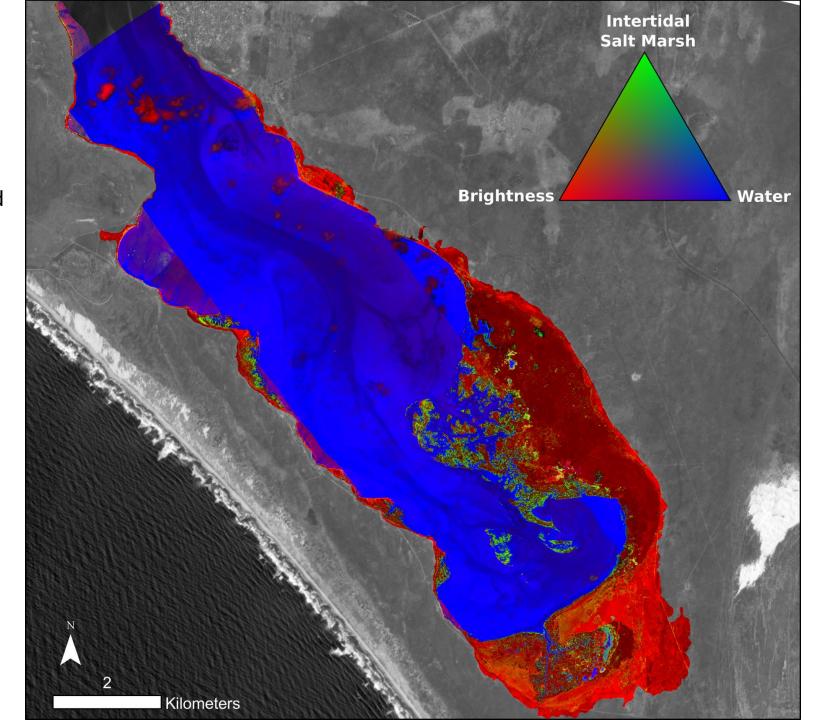
Langebaan Lagoon – 10/29/2023 (0.077 m above MSL)

- Brightness is used to normalize the MESMA results
- Tidal height difference derived from LVIS – 1.46 m

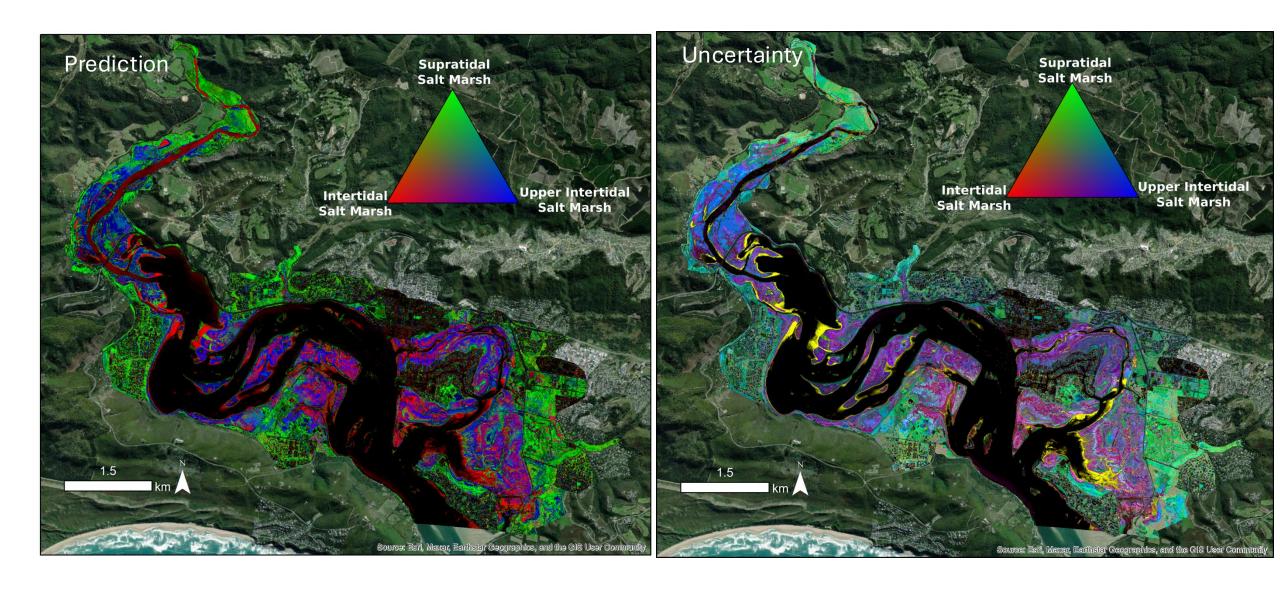




Co-associated habitat and materials.

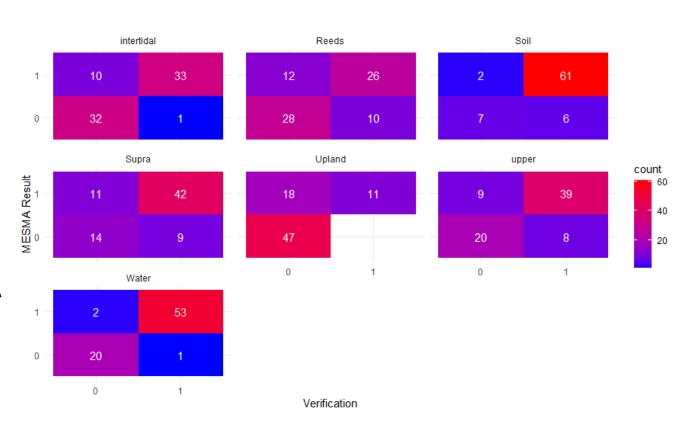


Knysna – MESMA result



MESMA-Plot verification

- MESMA result was extracted to plot locations and associated vegetation surveys were used to verify the absence or occurrence of each endmember.
- Of our 64 plots 1 plot was removed as it was outside of the estuarine functional zone and several plots had multiple MESMA results due to overlapping flight lines (n =76).
- Plots evaluated for each of our 8 classes.
- Overall accuracy: 81.4% (n=532)

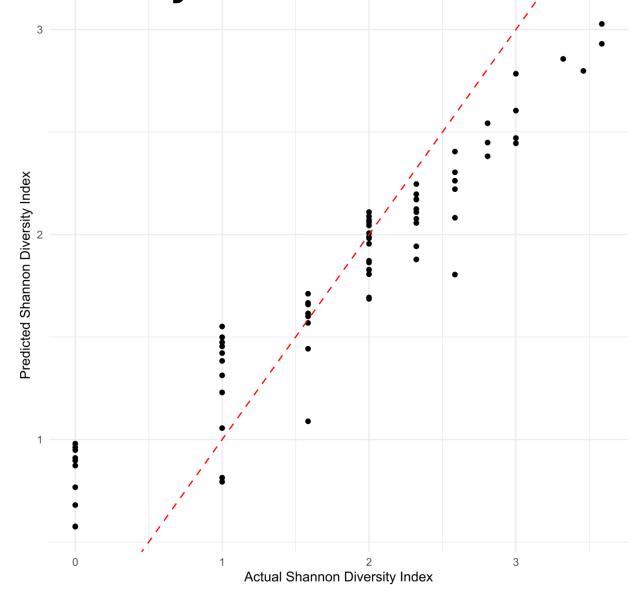


Predicting Shannon Diversity Index

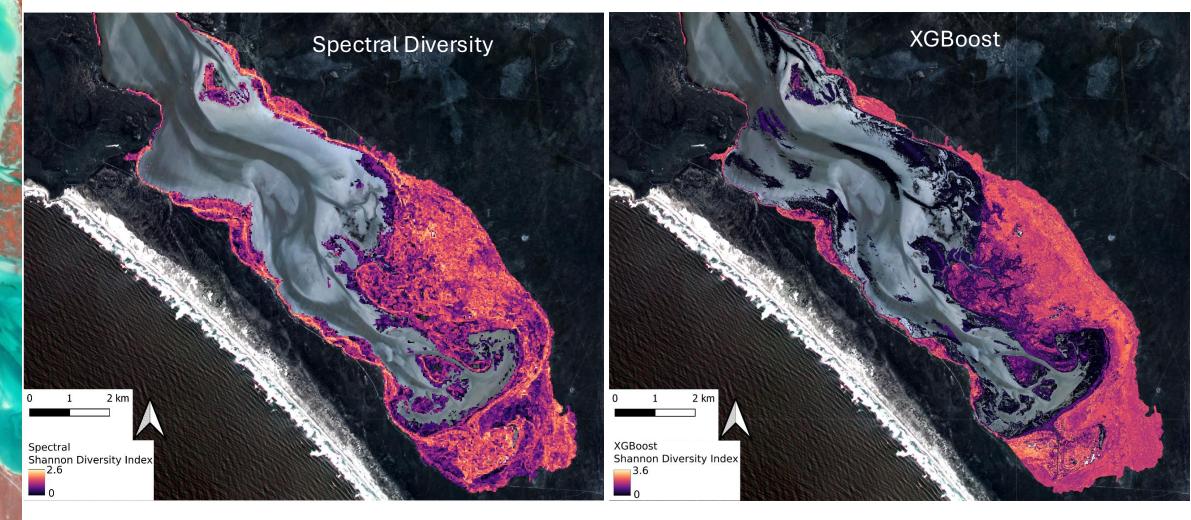
 Spectral diversity (Féret and Asner, 2014) alone did not capture the trend in Shannon diversity index seen in our plots

 ML algorithms did a good job but XGBoost overfit

Model	RMS E	R ²	MAE
Spectral Diversity	1.14	0.00 15	0.97
Random forest	0.42	0.91	0.32
Mean by Habitat (Salt marsh, SAV, Reeds & Sedges)	0.87	0.35	0.62
Multiple linear regression	0.78	0.30	0.617
XGBoost	0.12	0.99	0.098

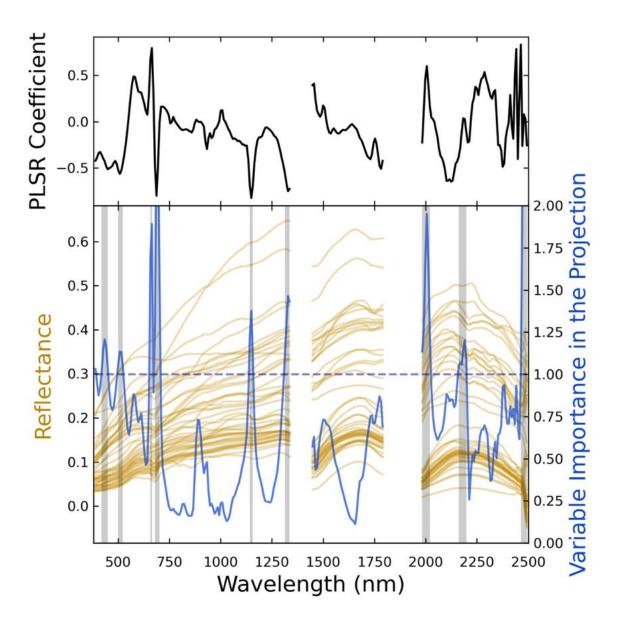


Shannon Diversity Index (Machine Learning vs Spectral)



Salinity

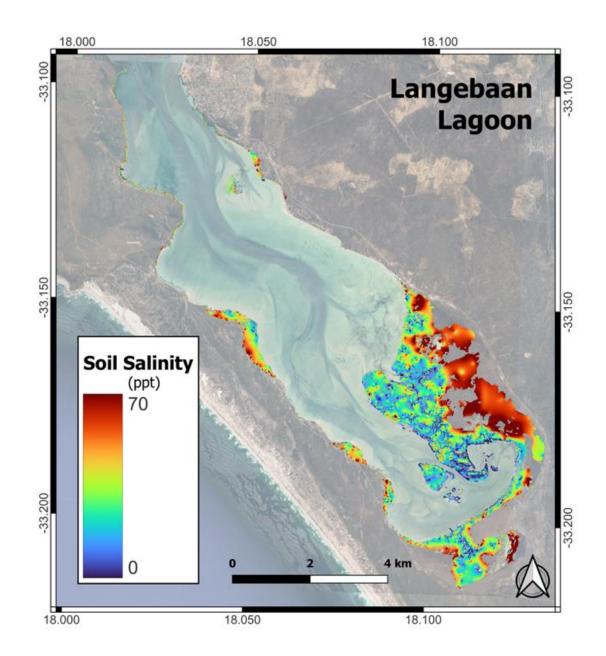
- Data collected in Carpenteria, CA, salt marsh was used to train a partial least squared regression of spectra to salinity.
- Langebaan Lagoon data collected as part of our fieldwork was used to validate the salinity prediction.



Salinity

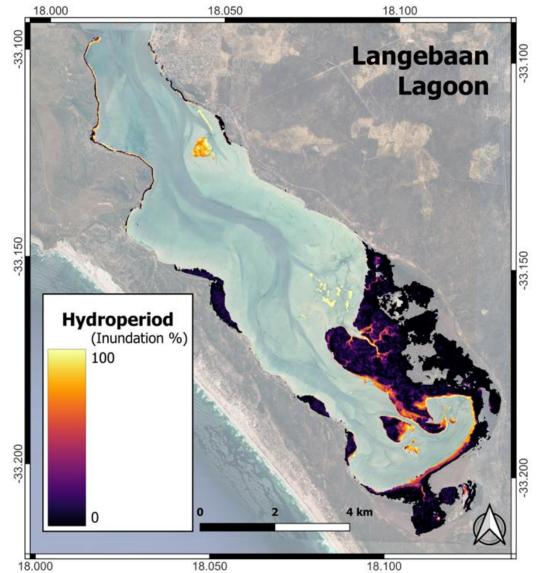
Soil salinity (parts per thousand) derived from AVIRIS-NG across the Langebaan Lagoon.

- Predicted for locations with a soil fraction > 0.8.
- bilinear interpolation was used to estimate soil in vegetated regions of the salt marsh.



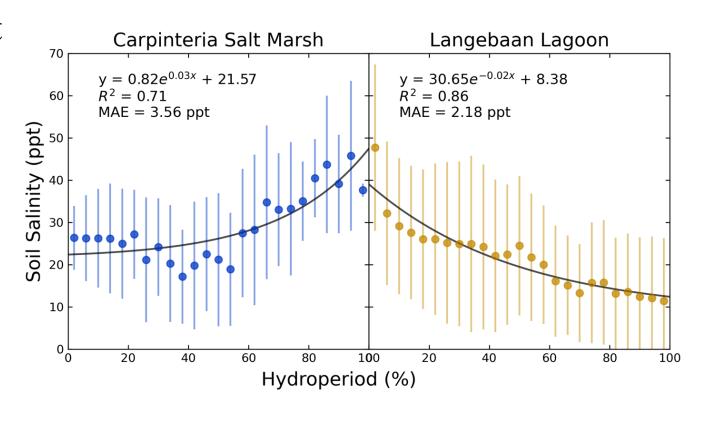
Hydrological parameters

- Hydroperiod- Sentinel-1 Synthetic Aperture Radar archive used to determine % of observations with inundation.
- Hydrological modeling integrating gauge data with bathymetry models to estimate the hydrological dynamics of Knysna estuary.





• Hydroperiod (% time spent inundated) derived from Sentinel-1 across (a) the Carpinteria Salt Marsh Reserve and (b) the Langebaan Lagoon. Each point represents the mean salinity value for the given bin, with the bars representing the bin's standard deviation.



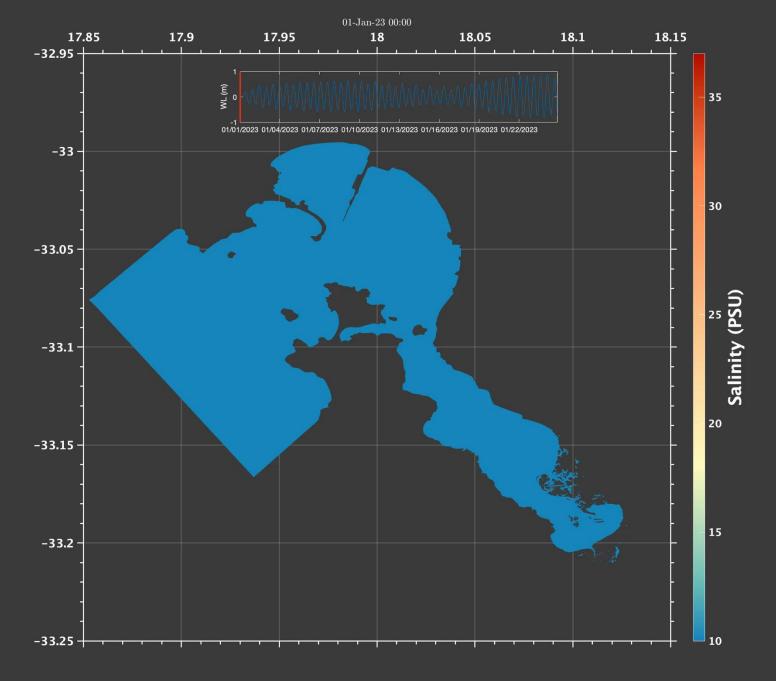
Jensen et al. *in prep*

Bathymetry- Langebaan Lagoon



Hydrodynamic Mesh





Conclusions and next steps

- Salinity prediction method performed well from California to South Africa
- MESMA class prediction are informative for calculate a spatial high resolution shannon diversity index that improved on Spectral Diversity metrics in our study estuaries.
- Next steps
 - Apply MESMA to all estuaries
 - Analyze estuarine drivers (salinity, elevation and hydroperiod) relative to Shannon diversity index

Acknowledgements









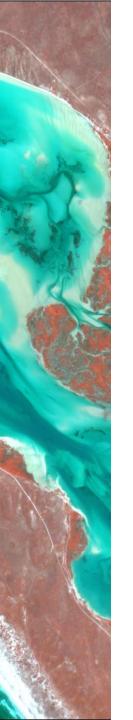












References

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