

Optilise kaugseire kasutamine merepõhja elupaikade kaardistamisel

Optical remote sensing for mapping seabed habitats

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Uuringu tellis ja uuringut rahastab Eesti Teadusagentuur ERF-ist programmi RITA kaudu. Uuring valmib Keskkonnaministeeriumi eesmärkide elluviimiseks



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Introduction

- Project activities:
 - 2.3.1. Optical remote sensing for mapping seabed habitats
 - 2.6.1. Mapping different features of fish spawning sites with remote sensing
- General aim: development of methodology of mapping shallow water seabed habitats (incl. fish spawning and nursery areas) using different optical remote sensing imagery (airborne orthophoto mosaics, hyperspectral imagery, satellite imagery) together with *in situ* sampling and mathematical modeling.

Substrate



Phytobenthos



Zoobenthos



Reefs



Sandbanks



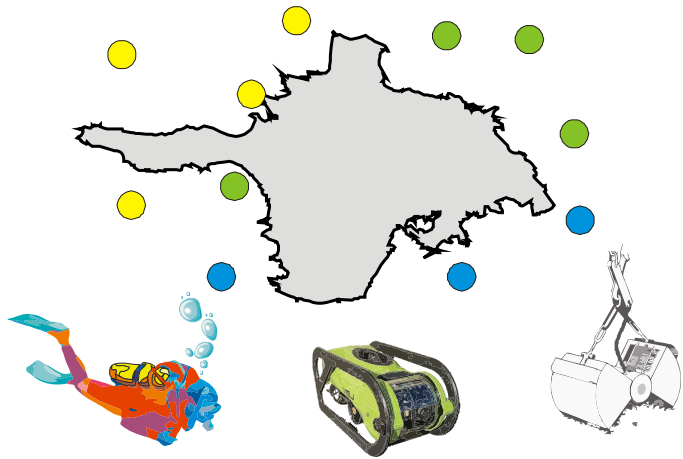
Why mapping of benthic habitats?

- Fundamental research
- Applied research
 - Environmental impact assessment
 - Environmental monitoring
 - Maritime spatial planning
 - Marine protected areas
 - Fish spawning and nursery areas
- EU obligations
 - Habitats directive (92/43/EEC)
 - Marine strategy framework directive (2008/56/EC)
 - Maritime spatial planning directive (2014/89/EU)

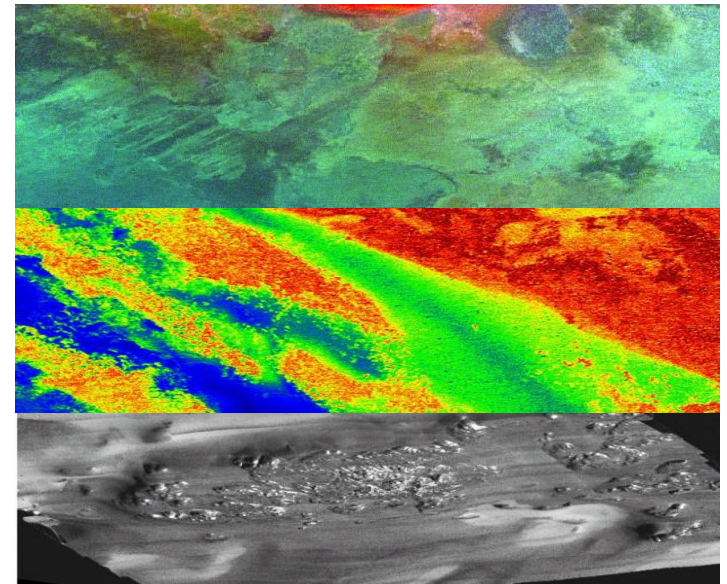
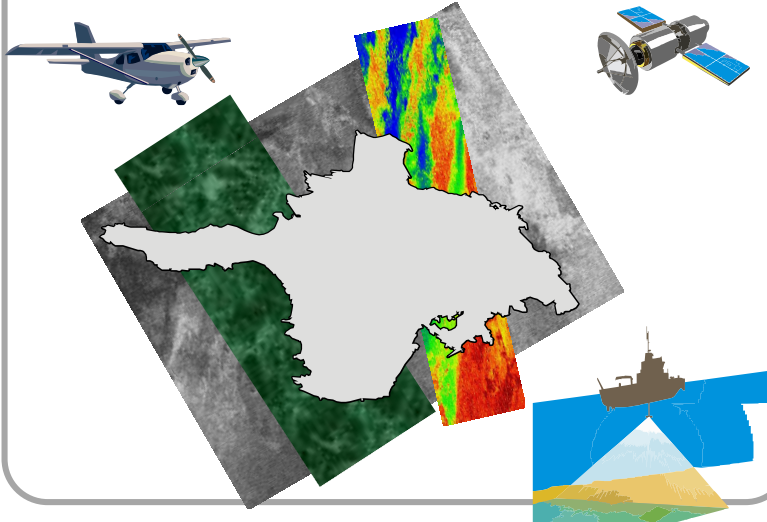
Fish spawning and nursery areas

- Potential species: pike, roach, ide, burbot
- Focus species – pike
- General preferences
 - Mainly spawning in fresh water
 - Dense vegetation (cover > 50%): hatched larvae attach to vegetation, later hiding
 - Water depth 0.2-1 m
- Preferences in fresh water
 - Flooded meadows/wetlands
 - Sedges, grasses, aquatic mosses, watermilfoils, charophytes, reed
 - Water depth 0.2-0.5 m
- Preferences in coastal areas
 - Dense vegetation
 - Common reed, charophytes, aquatic mosses, bladder wrack?
 - Water depth 0.2-1 m, several meters in open water
 - Freshwater influence

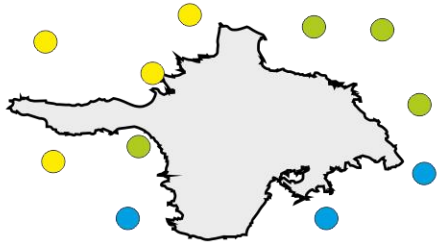
In situ research: substrate, biota



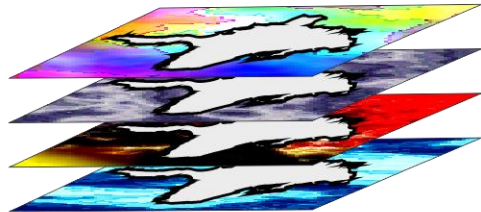
Remote sensing: optical, acoustic reflectance



Response variable
point data of species



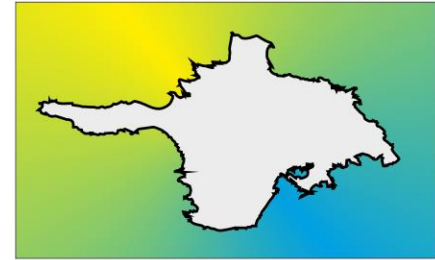
Predictor variables
GIS layers of
environmental data



Model
GAM
BRT
RF

Prediction

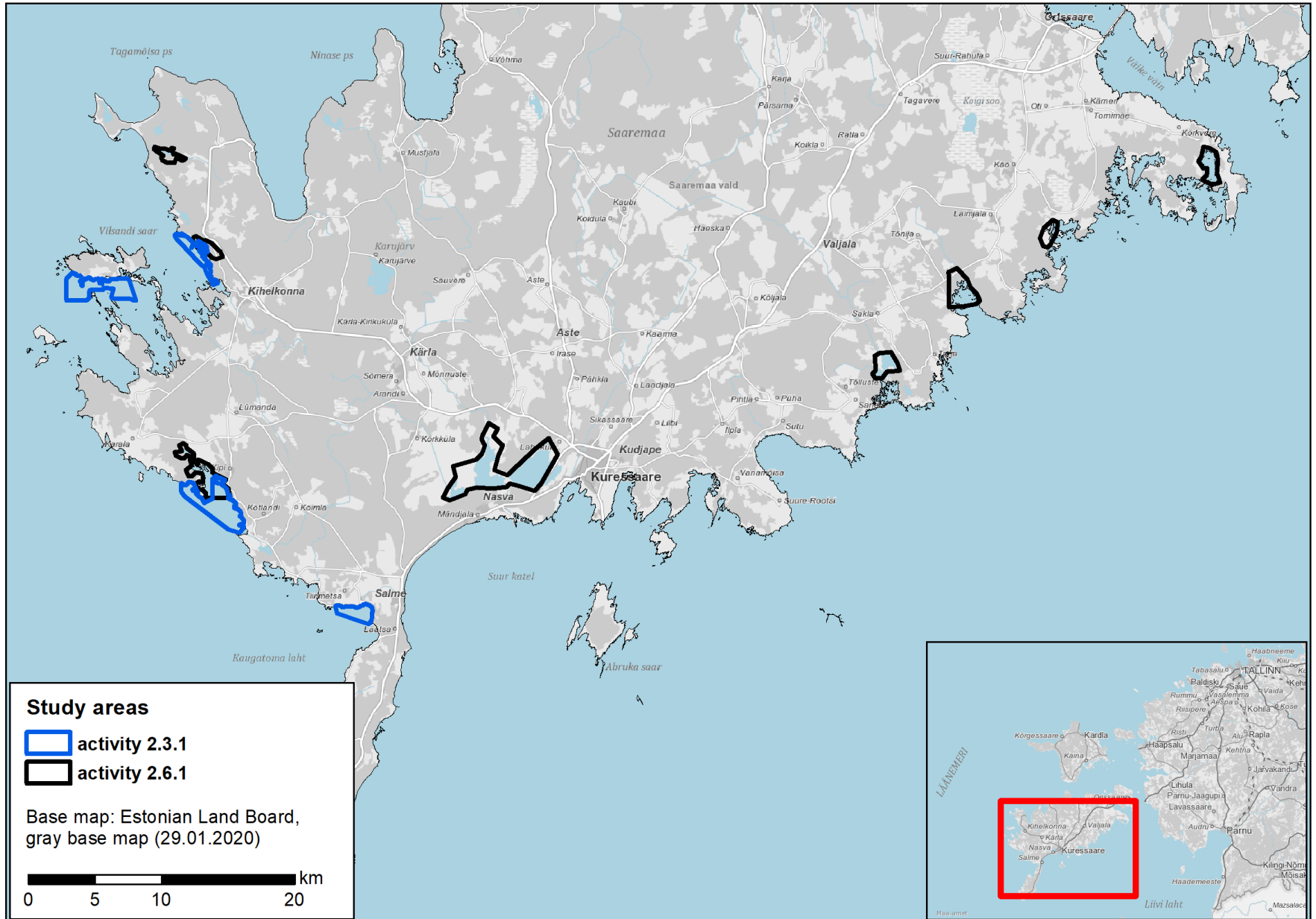
occurrence of species



Model assessment

- validation (model accuracy)
- importance of predictions
- shape of relationships

Material & methods



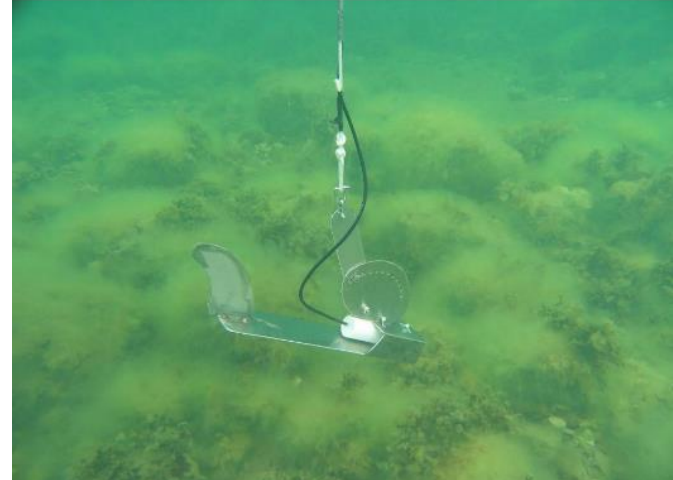
Remote sensing data

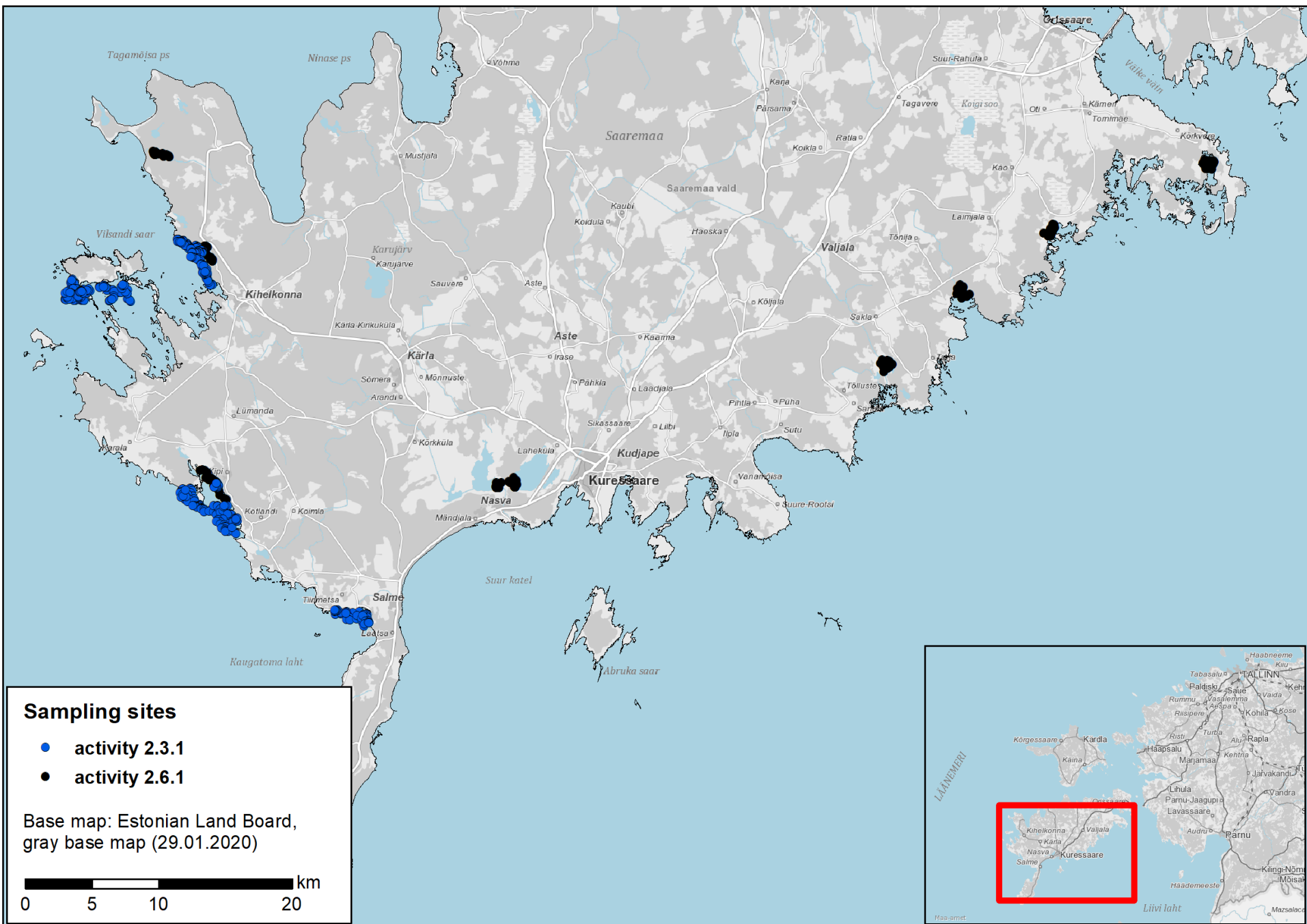
- Ortophotos of Estonian Land Board (LB)
 - RGB & NIR
 - 20 cm spatial resolution
 - 2017, 2019
 - Sun glint removal (Hedley et al 2005)
 - Model input: R, G, B, R/G, G/B, R/B
- Drone orthophotos
 - DJI Phantom 4 Pro V2.0
 - RGB
 - 3-6 cm spatial resolution
 - June, July 2019
 - Model input: R, G, B, R/G, G/B, R/B

- Hyperspectral imagery
 - Hyperspectral imager Hypspx (Norsk Elektro Optikk)
 - TU EMI's instrument on Land Board's airplane
 - 409-988 nm, 2.7 nm spectral resolution, 216 bands
 - 80 cm spatial resolution
 - July 2019
 - Model input:
 - Selected 12 bands
 - 12 first PCA components
- Sentinel-2 satellite imagery
 - 443-740 nm, 6 bands
 - 10-20 m spatial resolution
 - June, July 2019
 - Model input: 6 bands

***In situ* sampling**

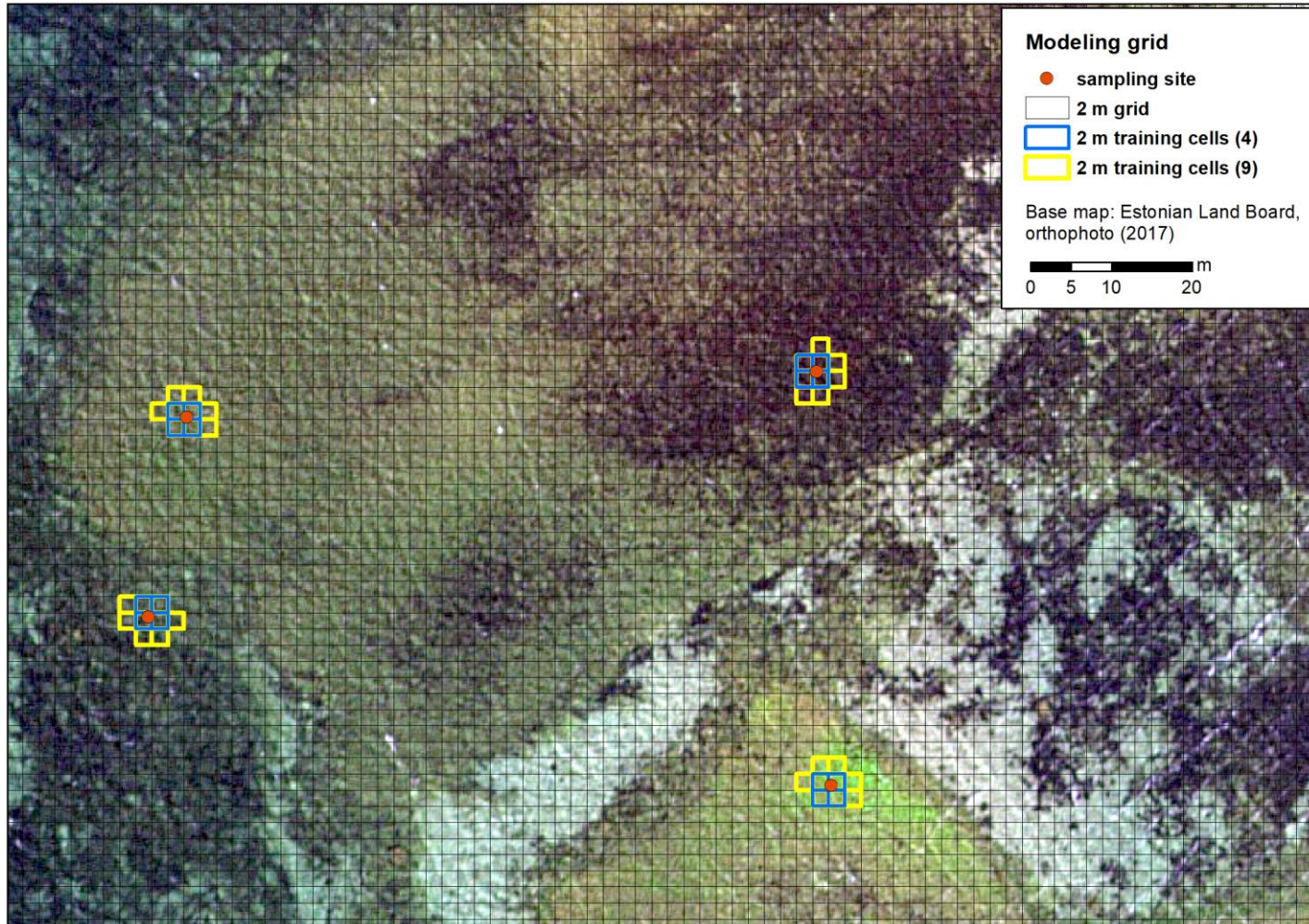
- 500 sites in 2.3.1
- 176 sites in 2.6.1
- Underwater video, handheld camera, direct observations
- Percentage cover of seabed substrate types, macrophytes, epibenthic bivalves, barnacles





Compilation of training and prediction data sets

- 2 m modeling grid, mean and st. dev. in each cell
- LB orthophoto and Hypspx: 4 and 9 nearest cells to sampling site



- Drone orthophoto: manual selection of grid cells
- Satellite: pixel value of sampling site

Modeling

- Algorithms
 - Random forest (RF)
 - Boosted regression trees (BRT)
 - Neural networks (NN)

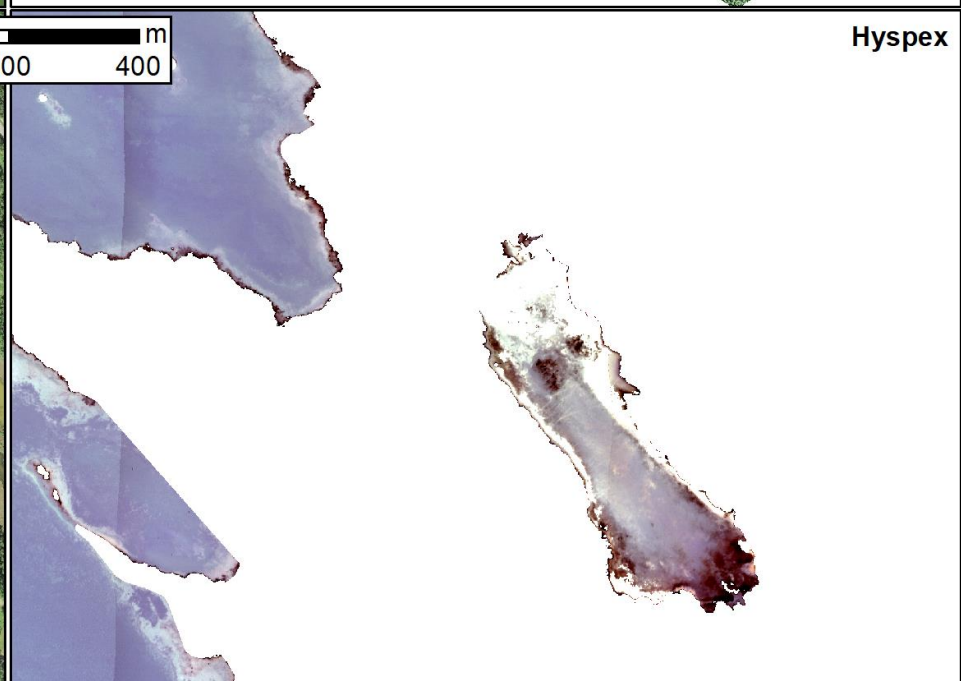
- Response variables
 - Seabed class (2.6.1)
 - Depth of water column (2.6.1)
 - Presence of reed belt (2.6.1)
 - Cover of hard substrate
 - Cover of sand
 - HELCOM HUB substrate type
 - Total cover of macrovegetation
 - Key species/groups: *Fucus*, *Zostera*, charophytes, filamentous algae, green algae, brown algae, red algae
- Additional predictor variables:
 - Depth
 - Wave exposure
 - Geographical coordinates

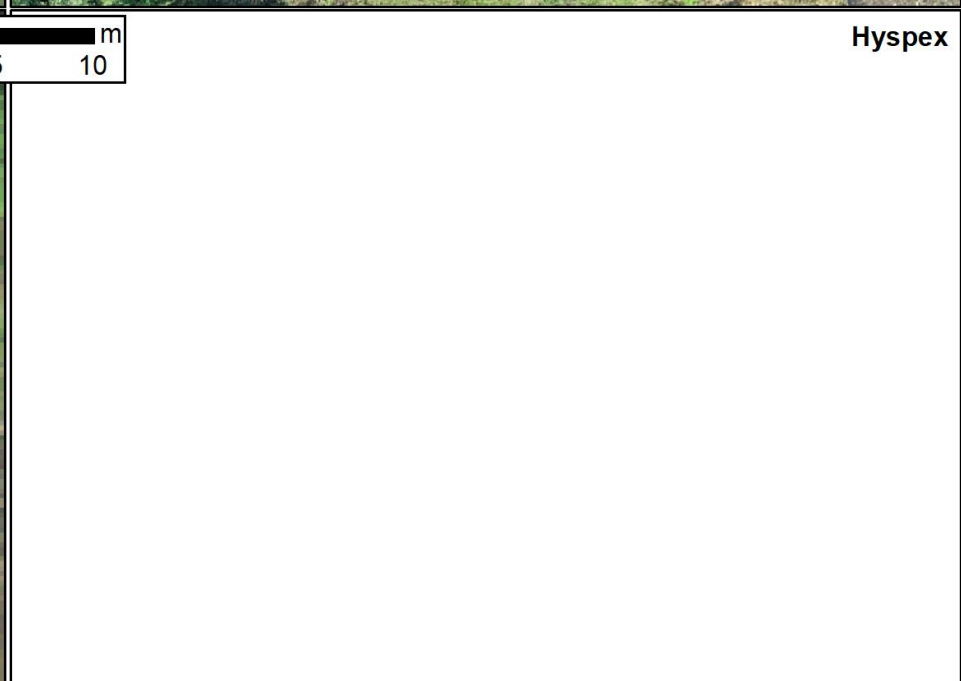
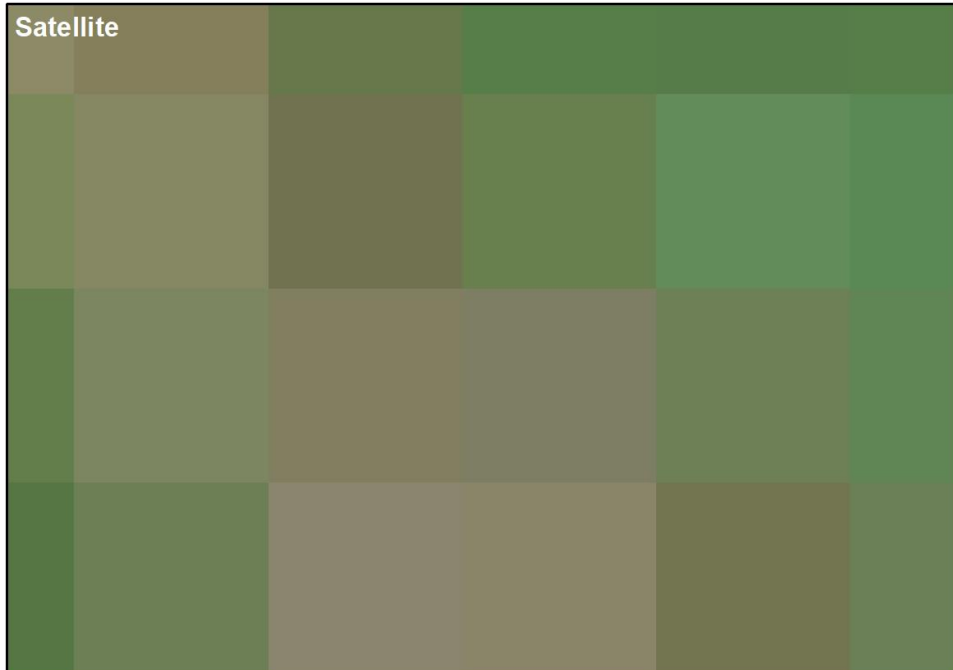
- Different sets of remote sensing predictors and additional predictors
- Area-based and pooled models
- R

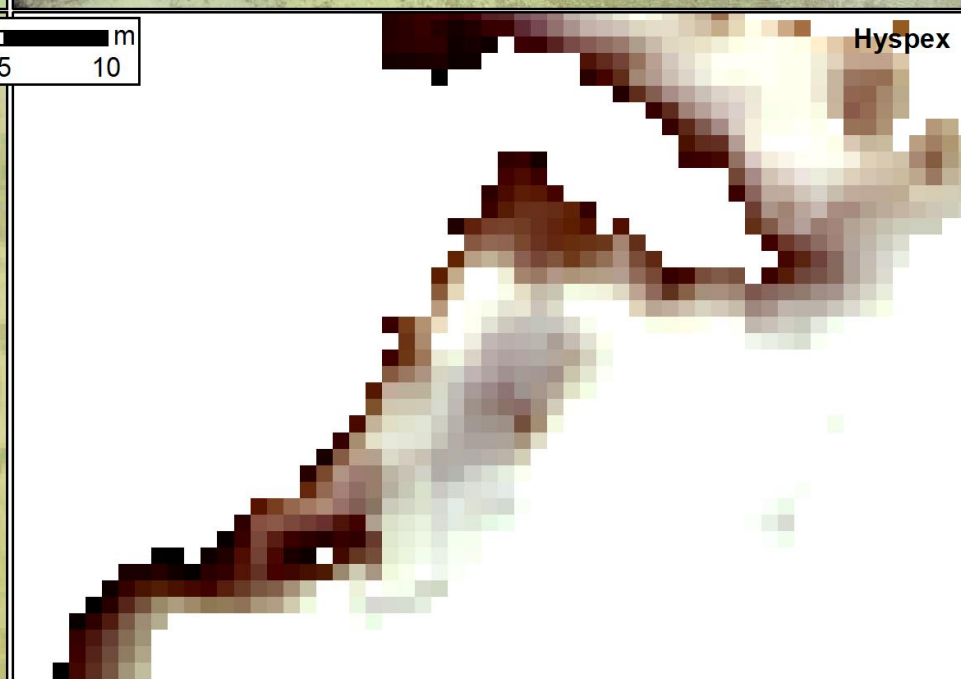
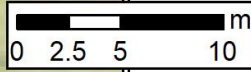
Model validation

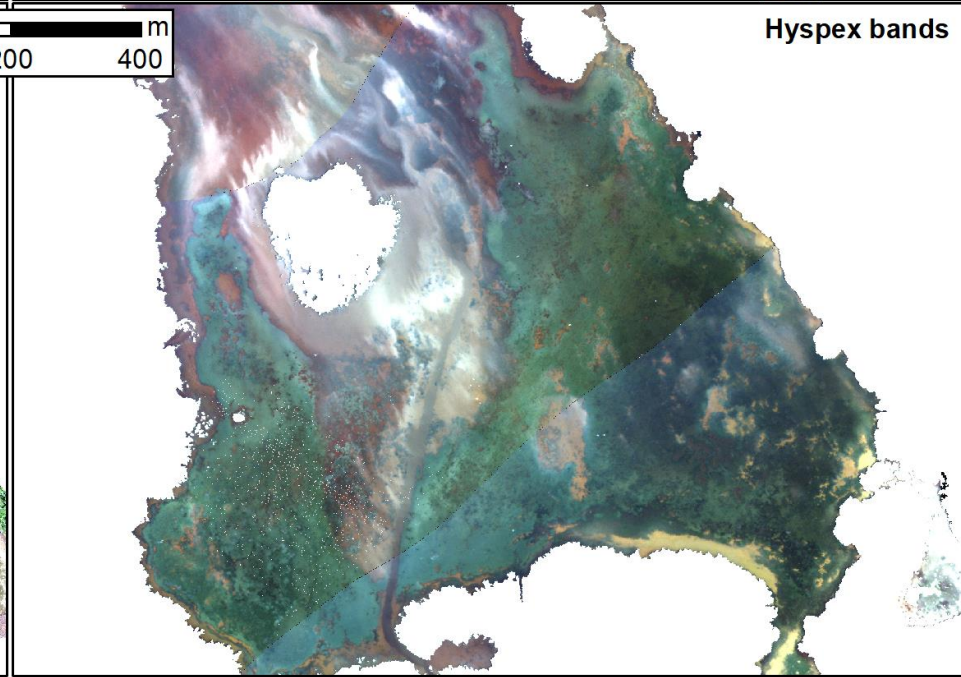
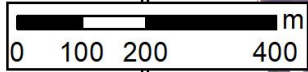
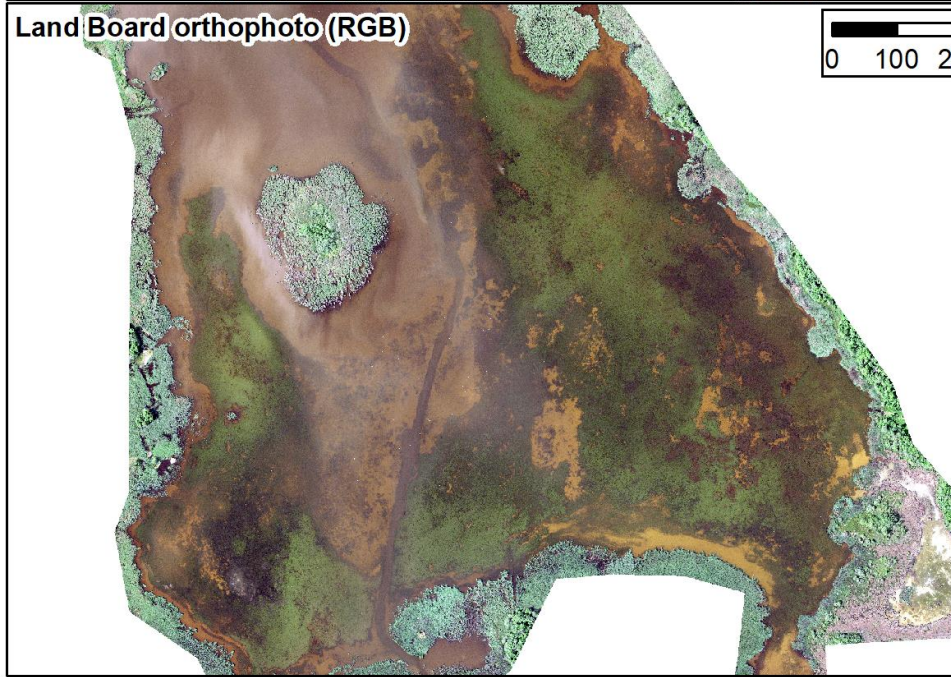
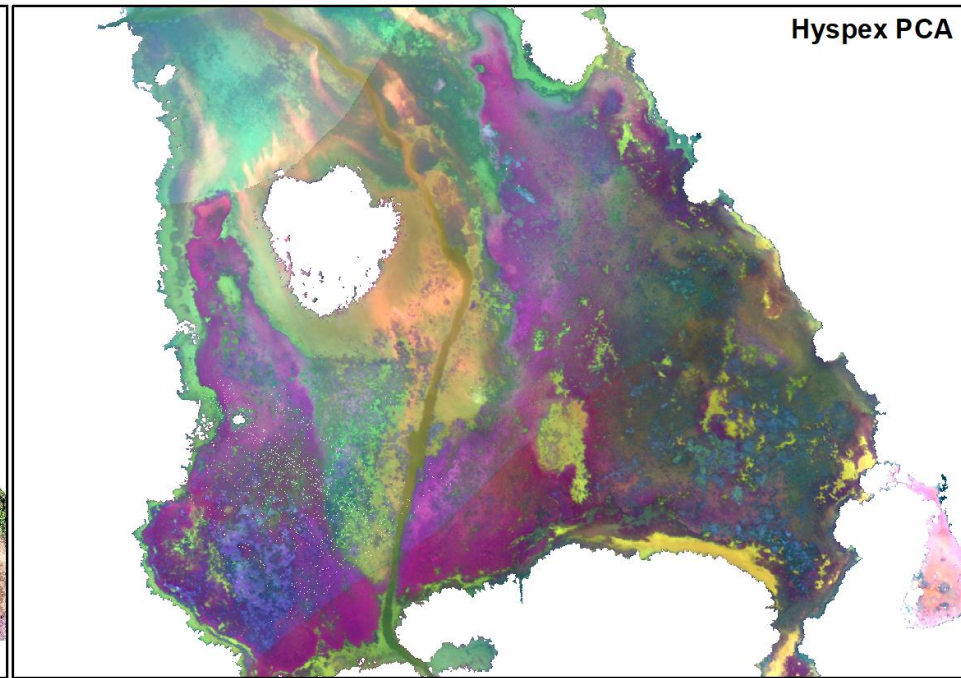
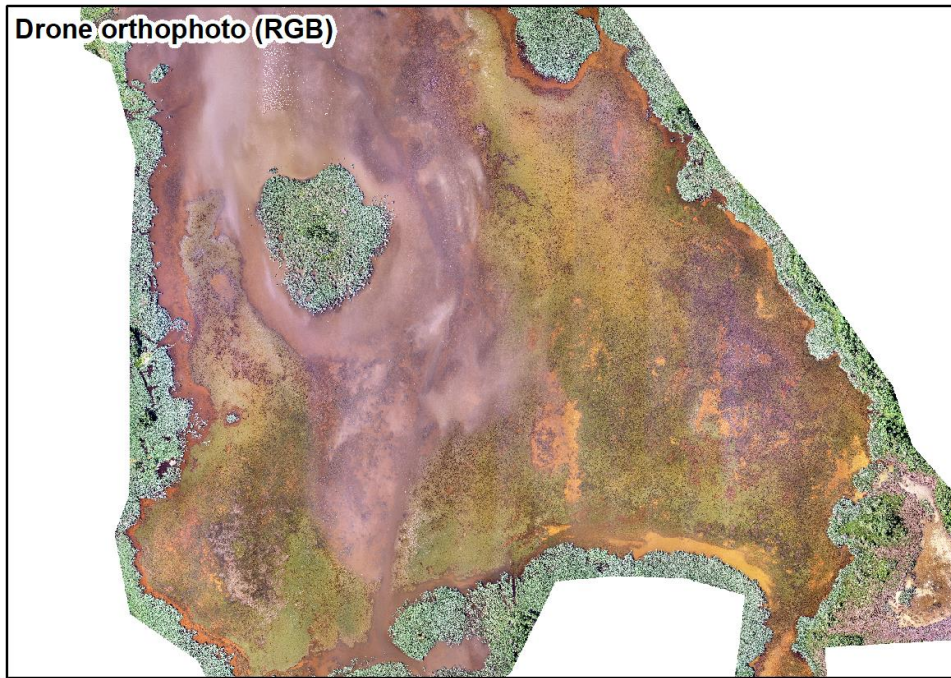
- Mathematical
 - 25% data reserved for external validation
 - 100% data for spatial predictions
 - Correlation coefficient, mean absolute error, classification accuracy, Kappa coefficient
- Visual expert assessment

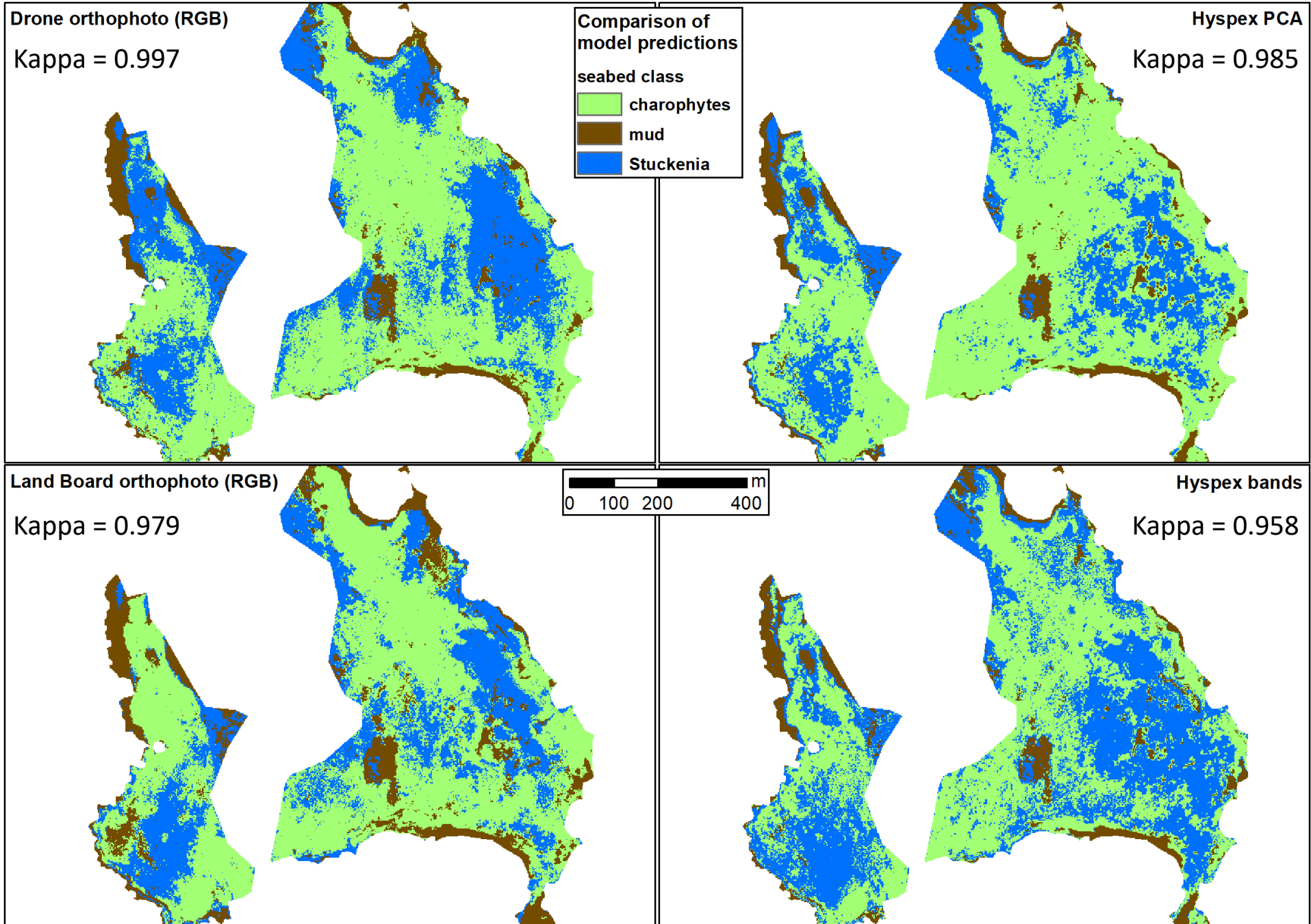
Results



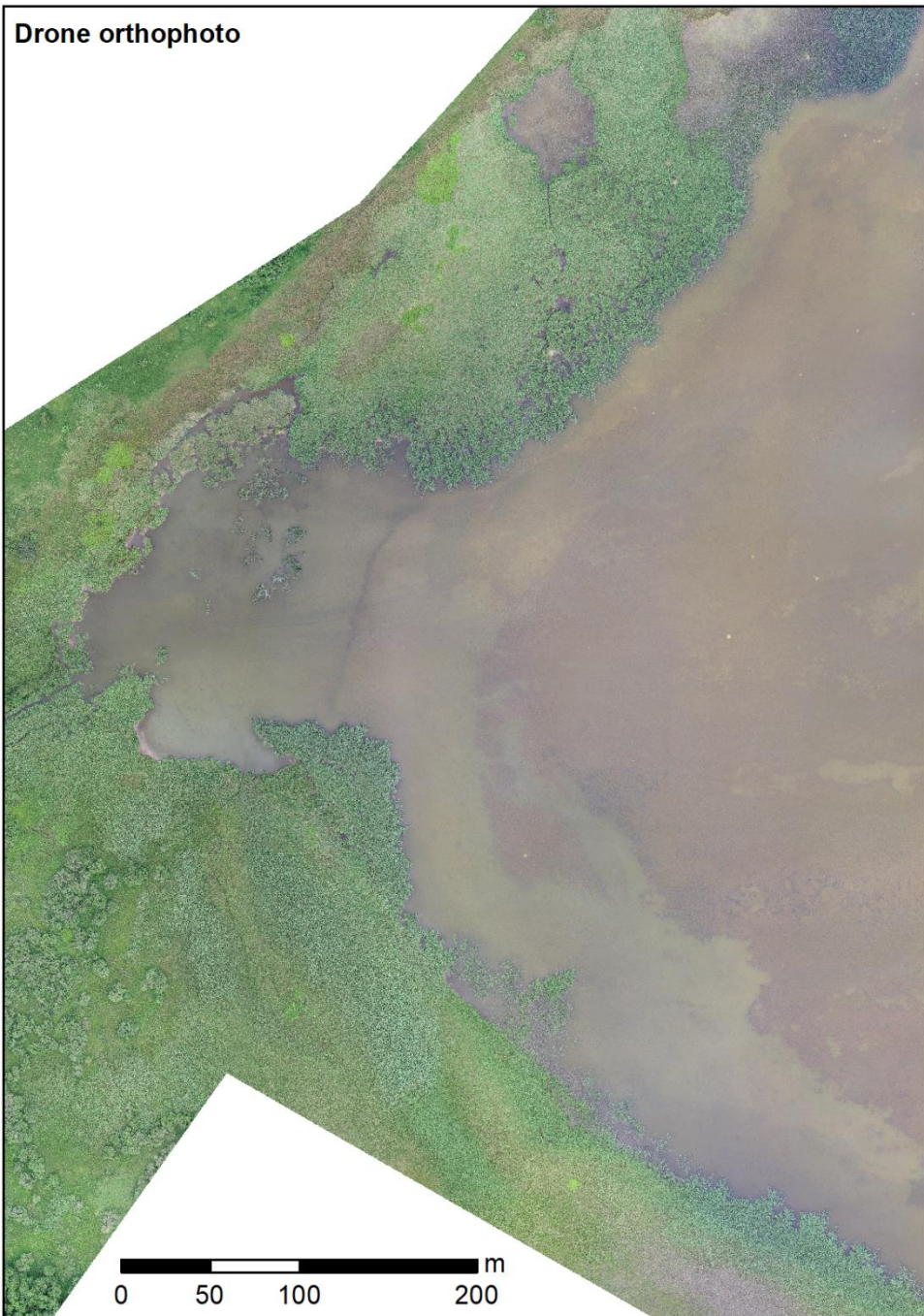




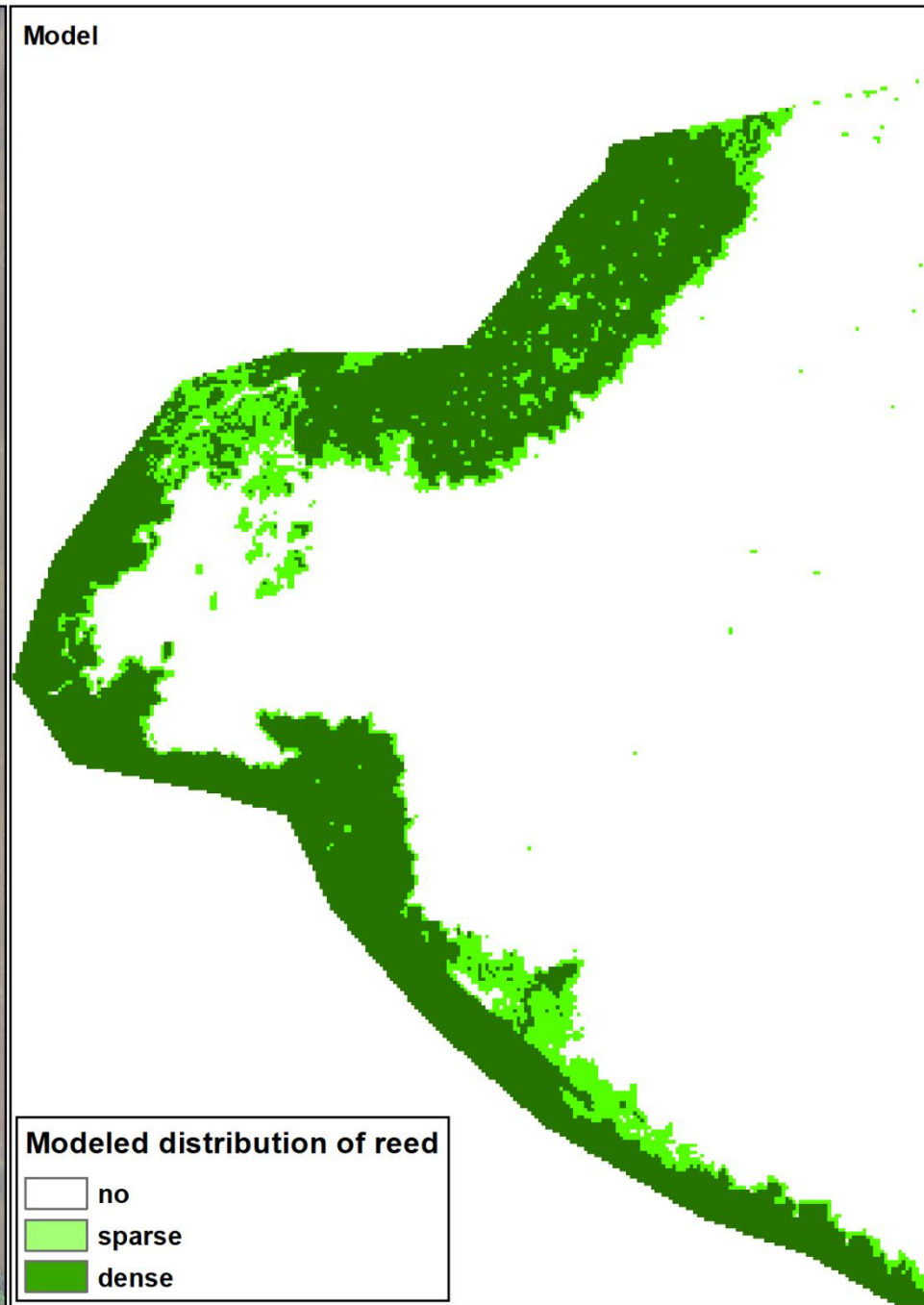


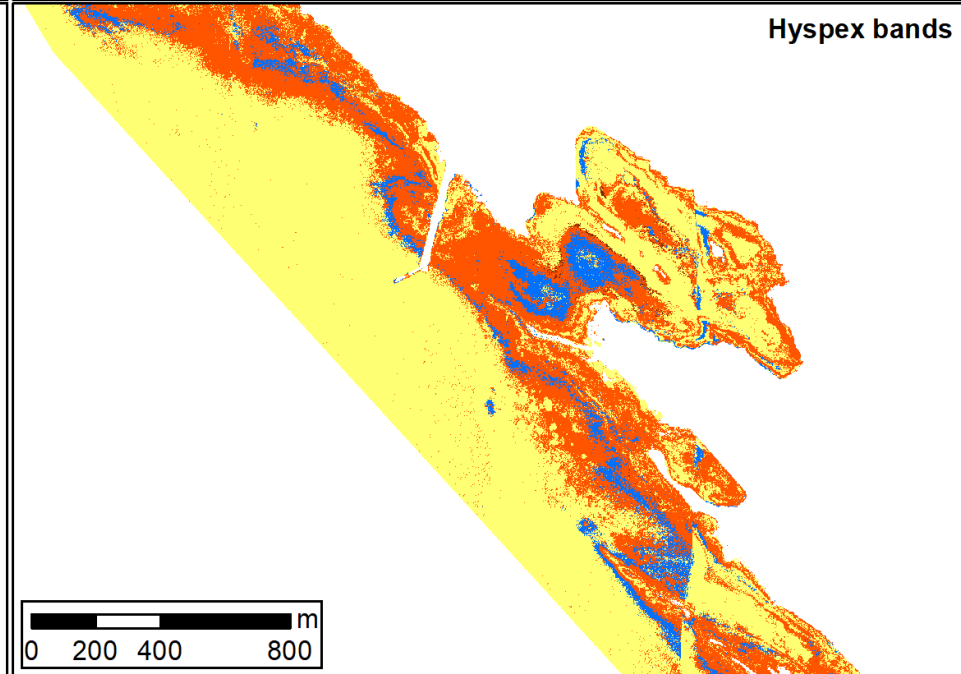
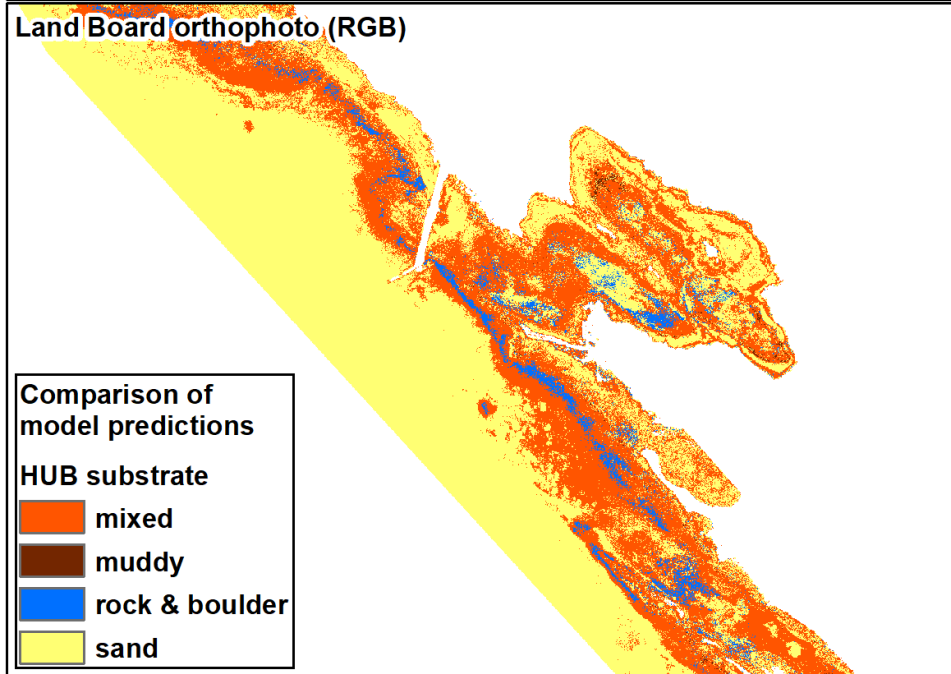
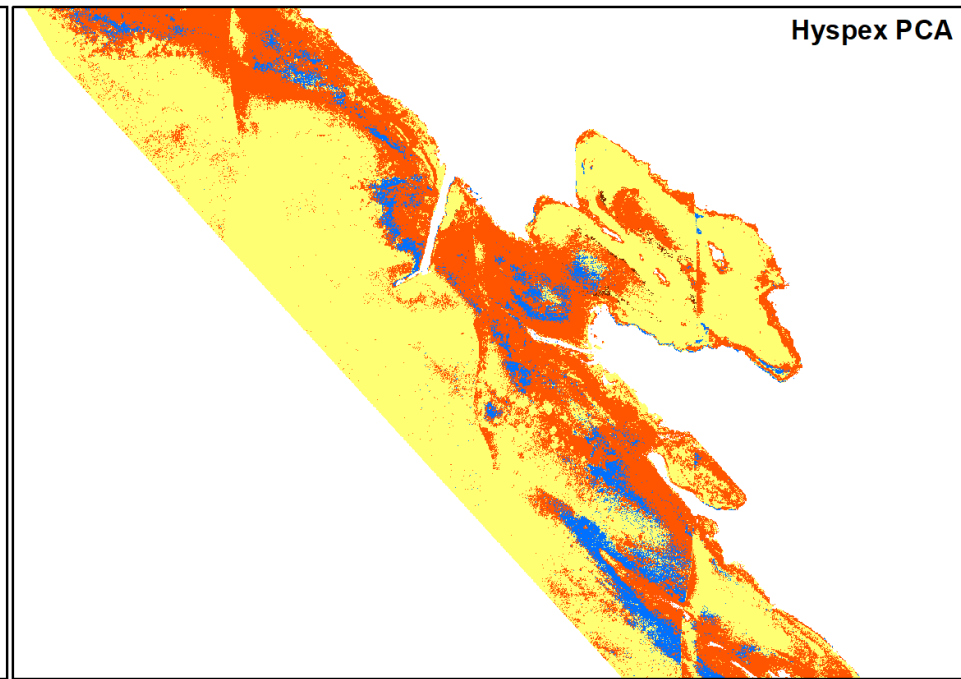
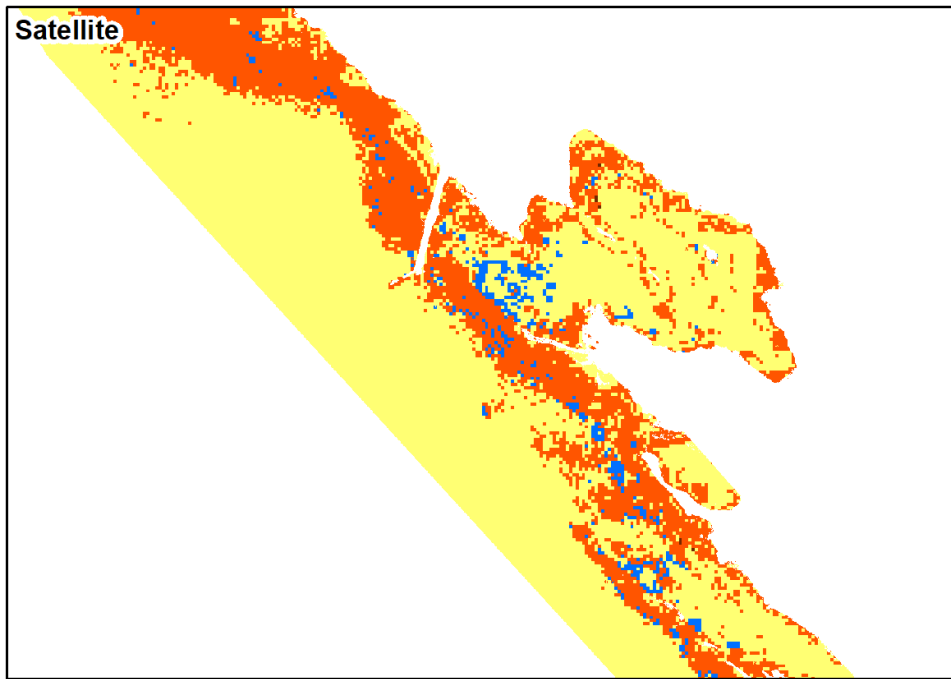


Drone orthophoto







Model

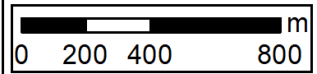


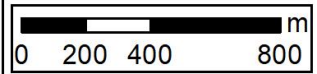
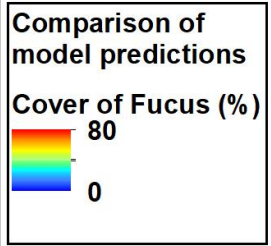
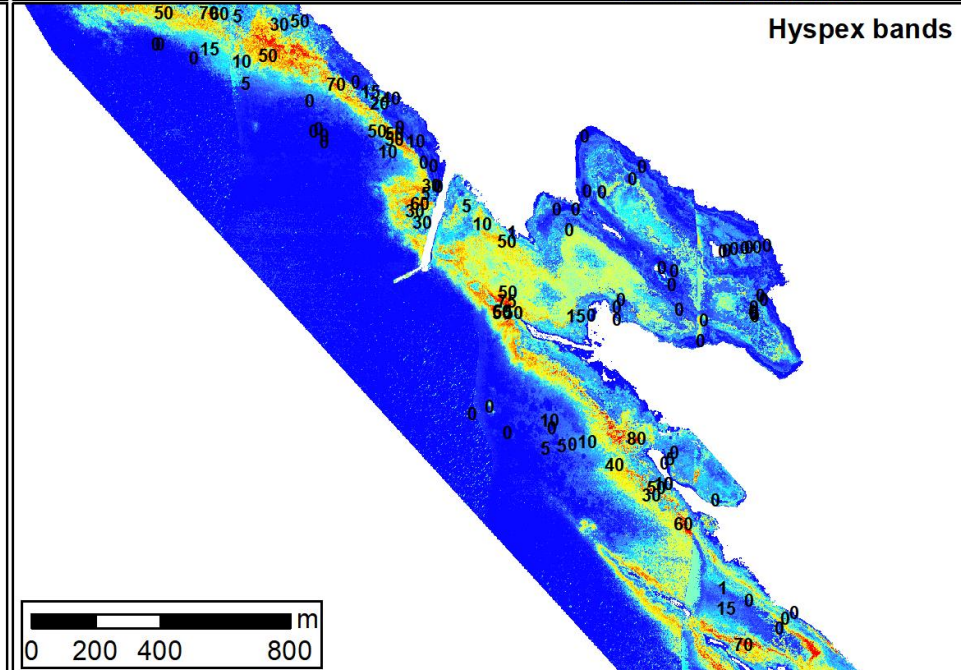
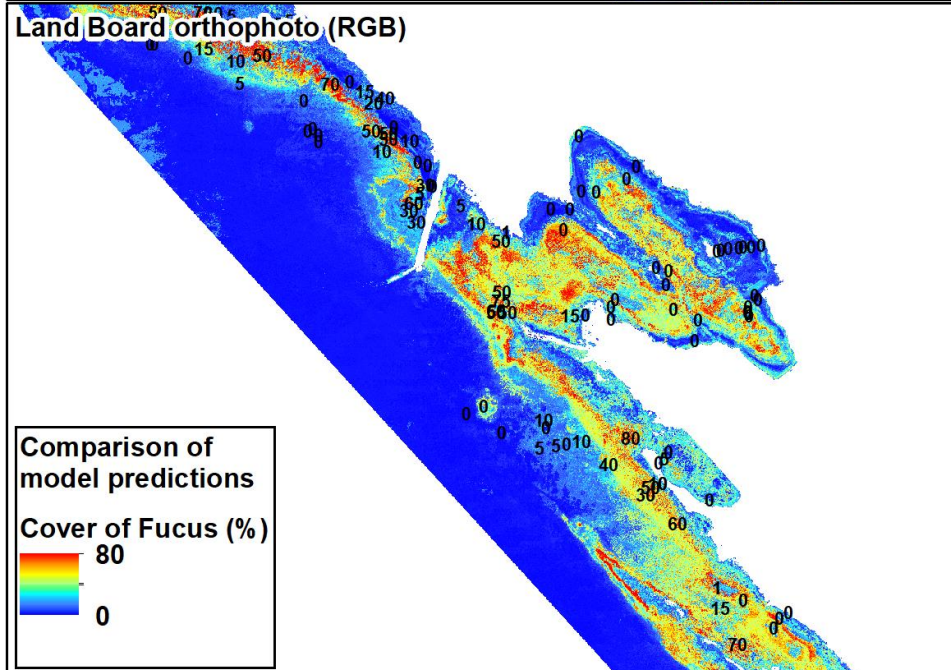
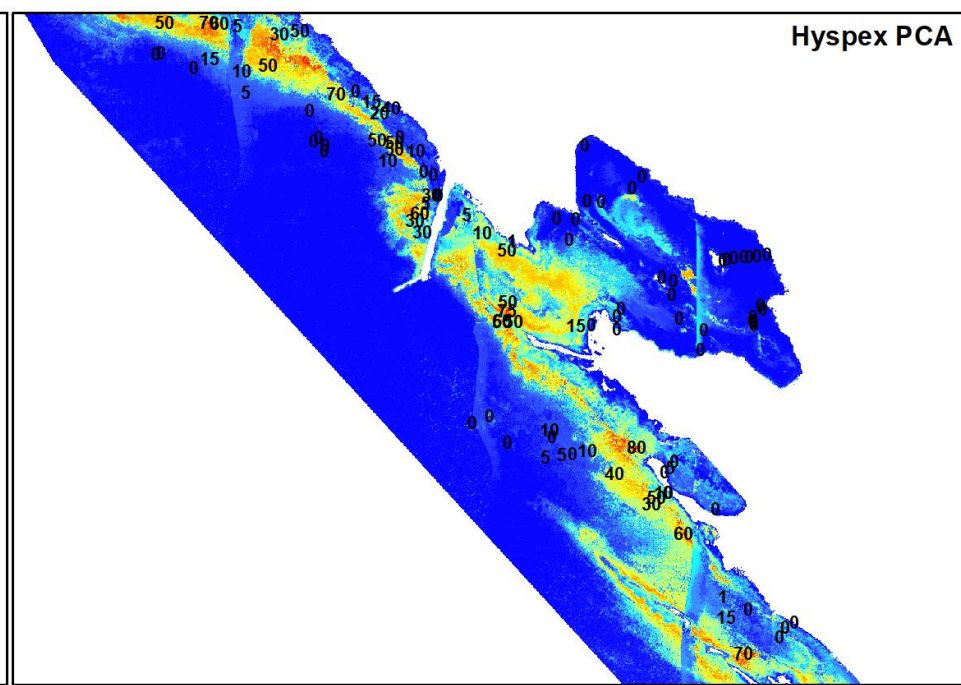
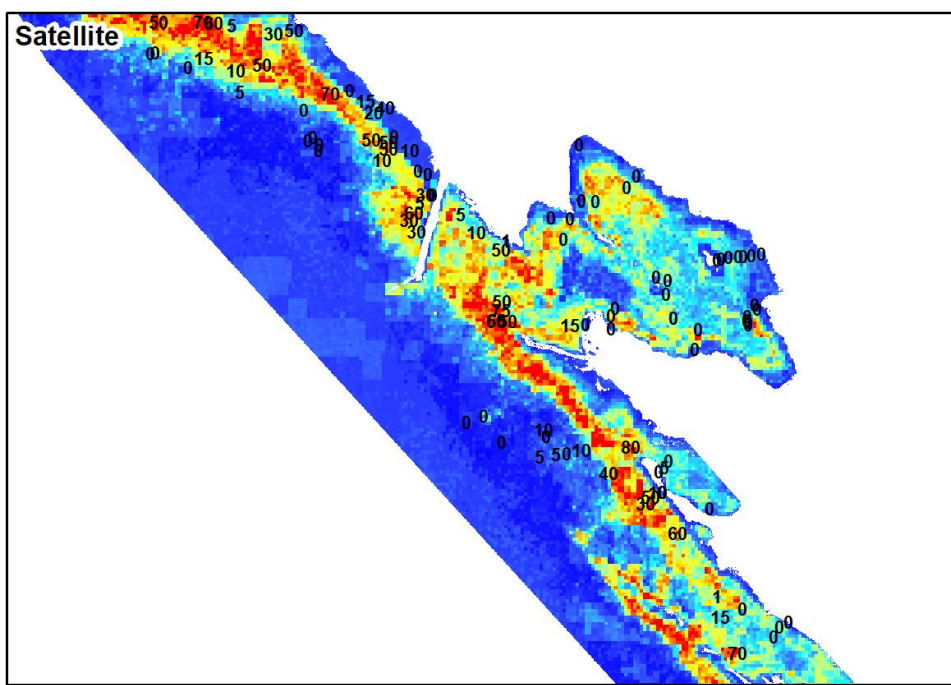


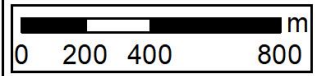
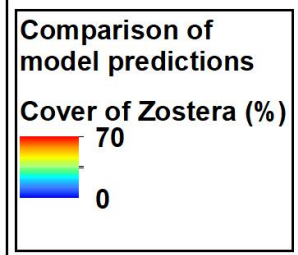
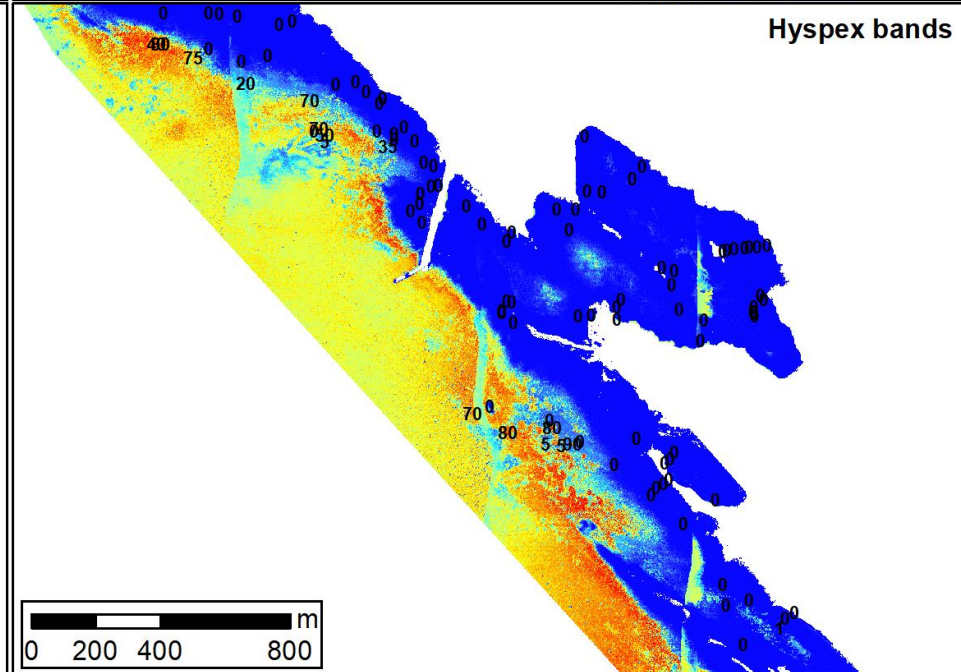
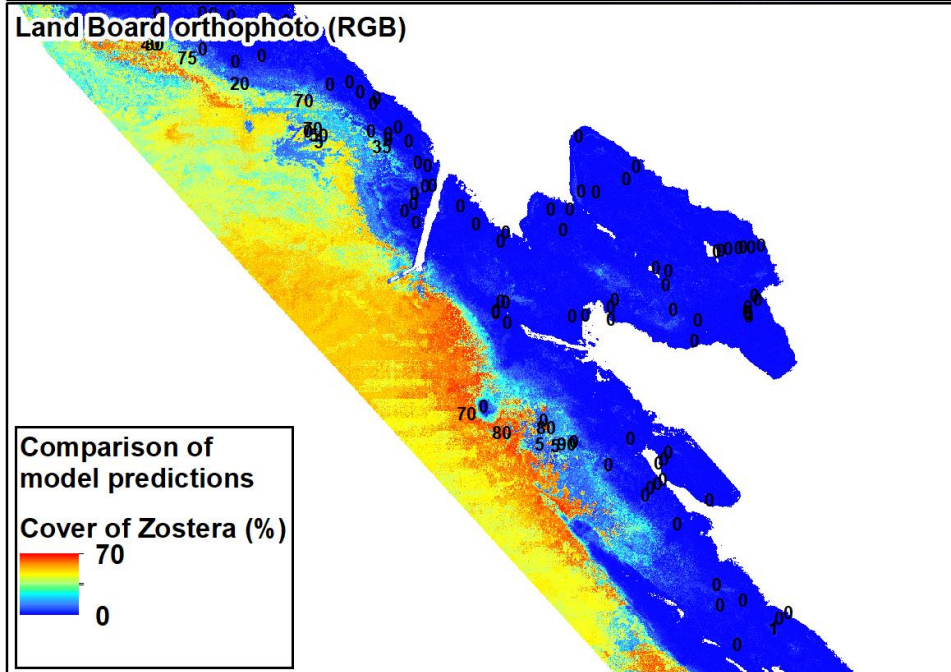
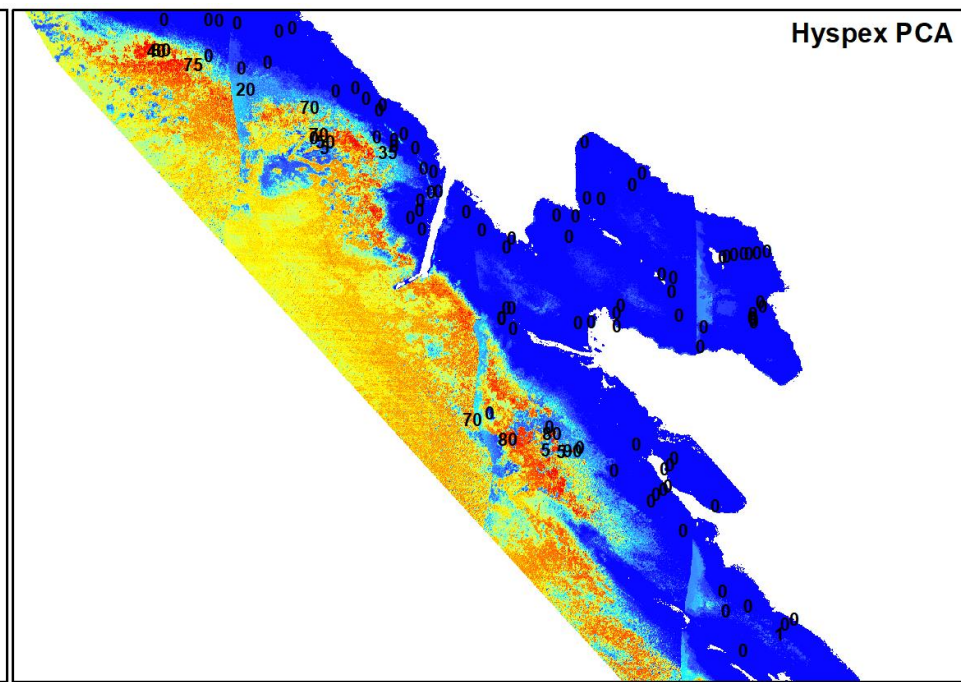
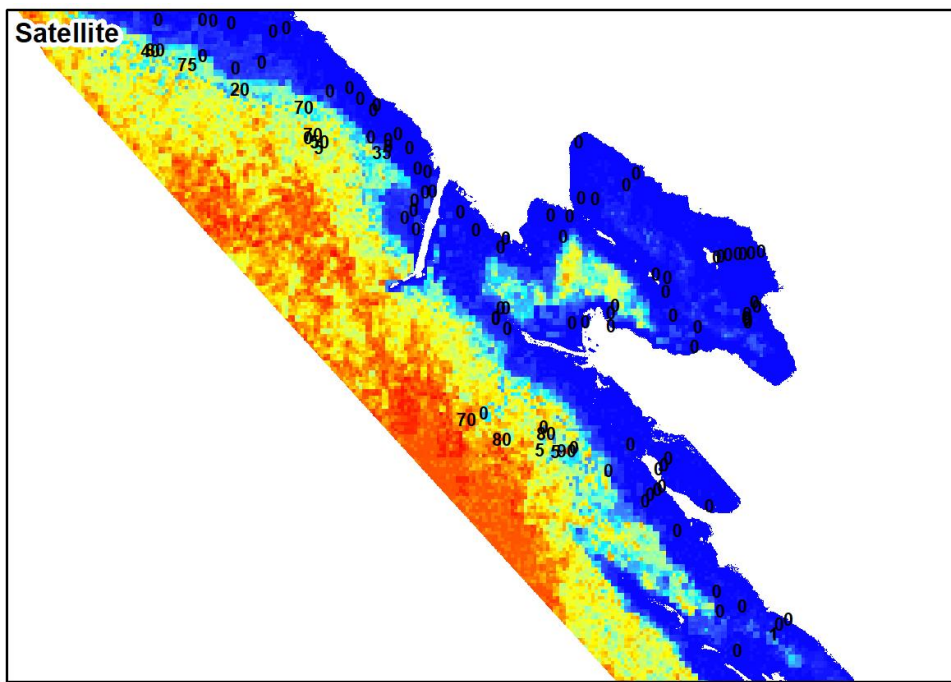
Comparison of model predictions

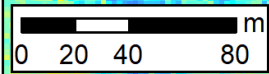
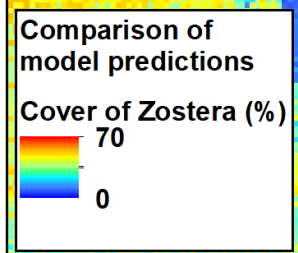
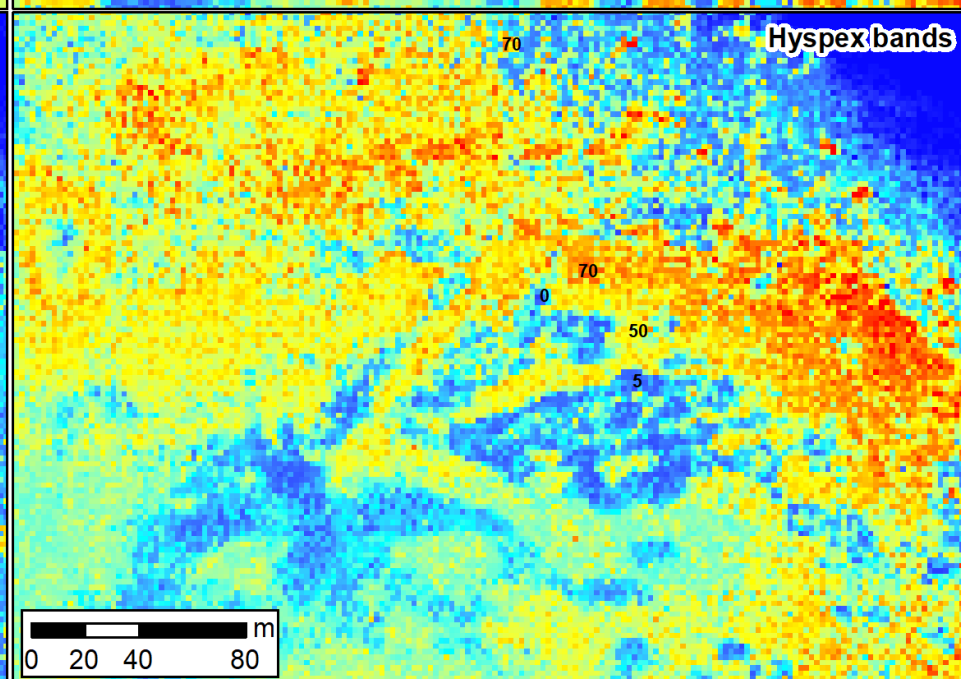
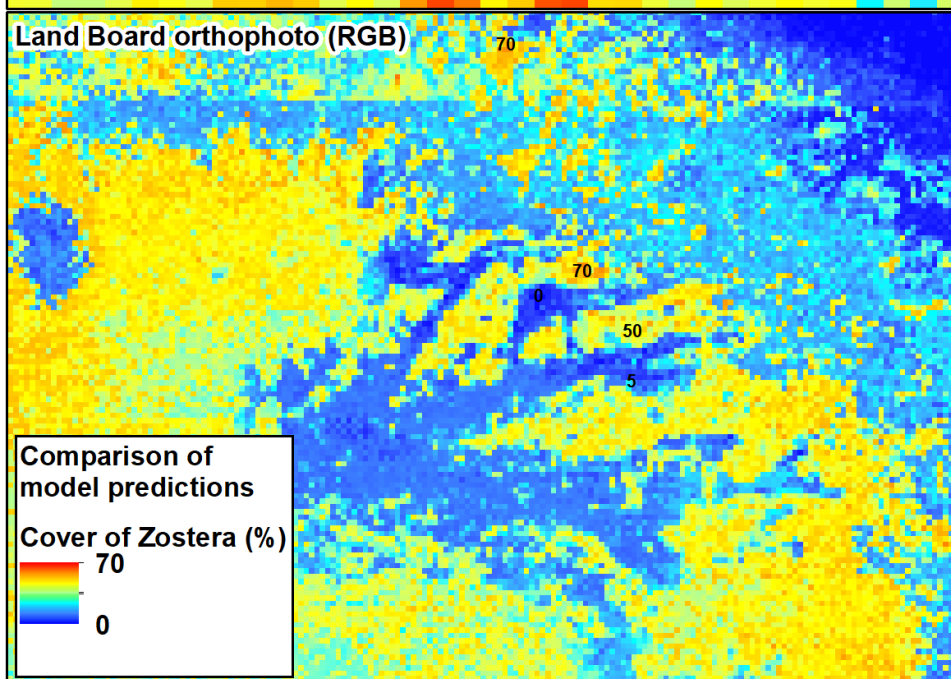
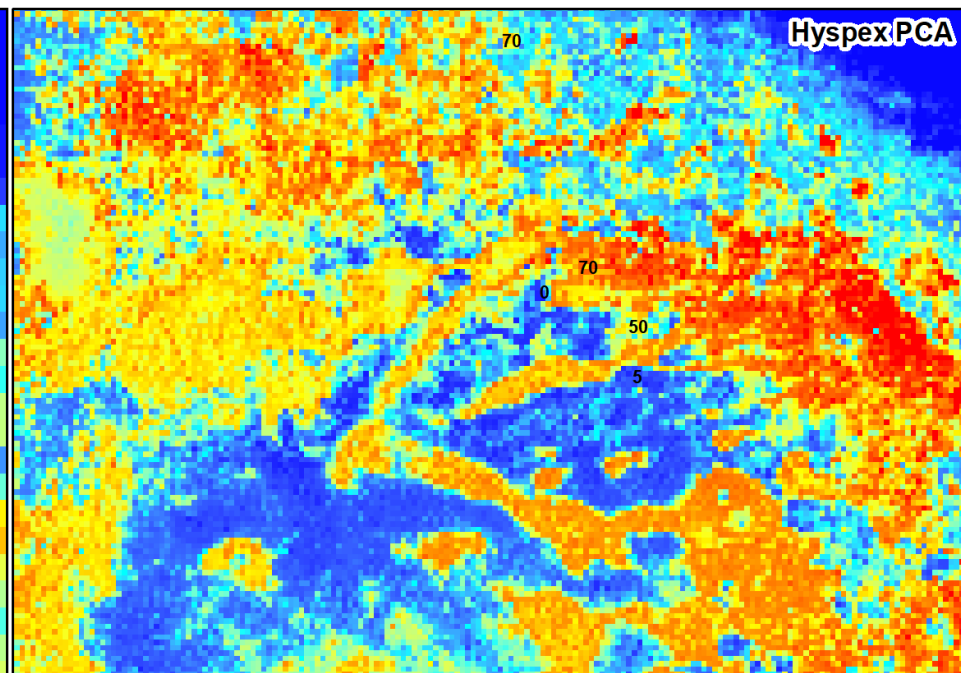
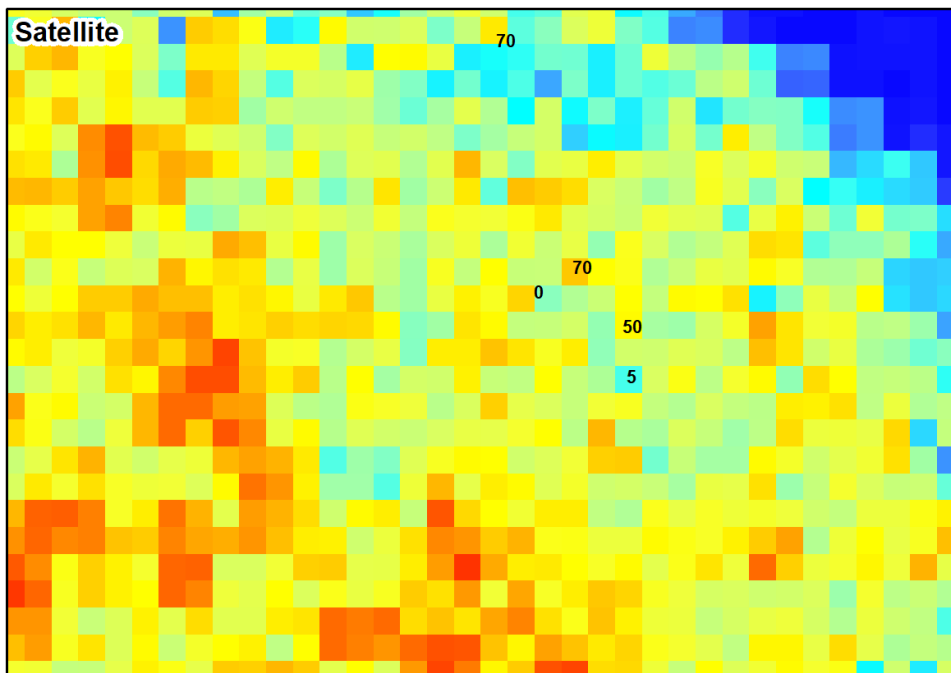
HUB substrate

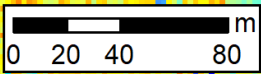
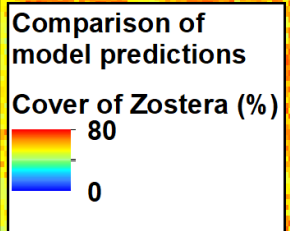
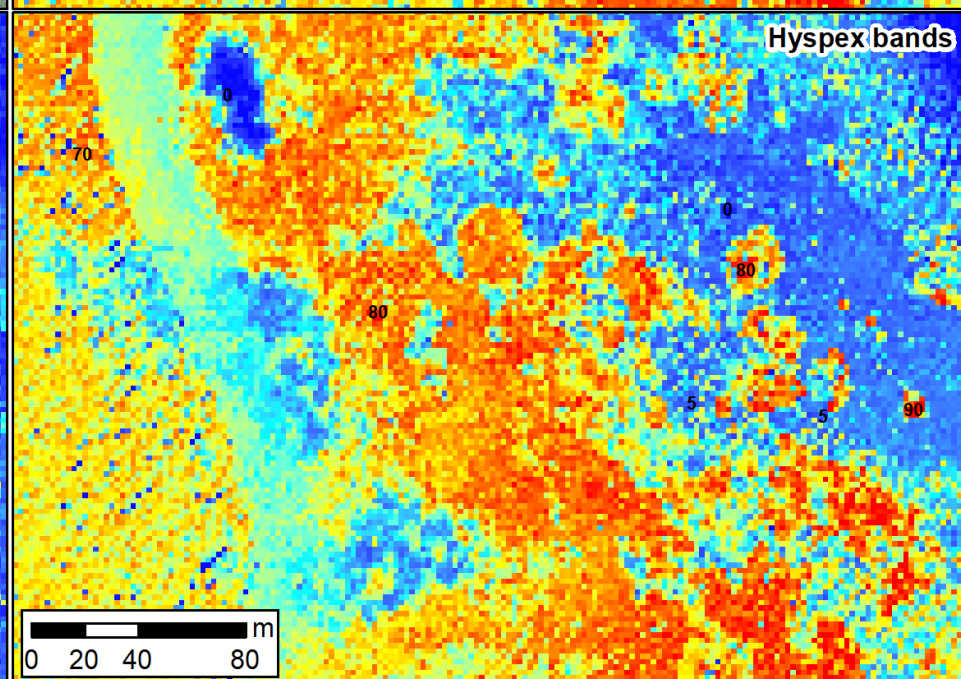
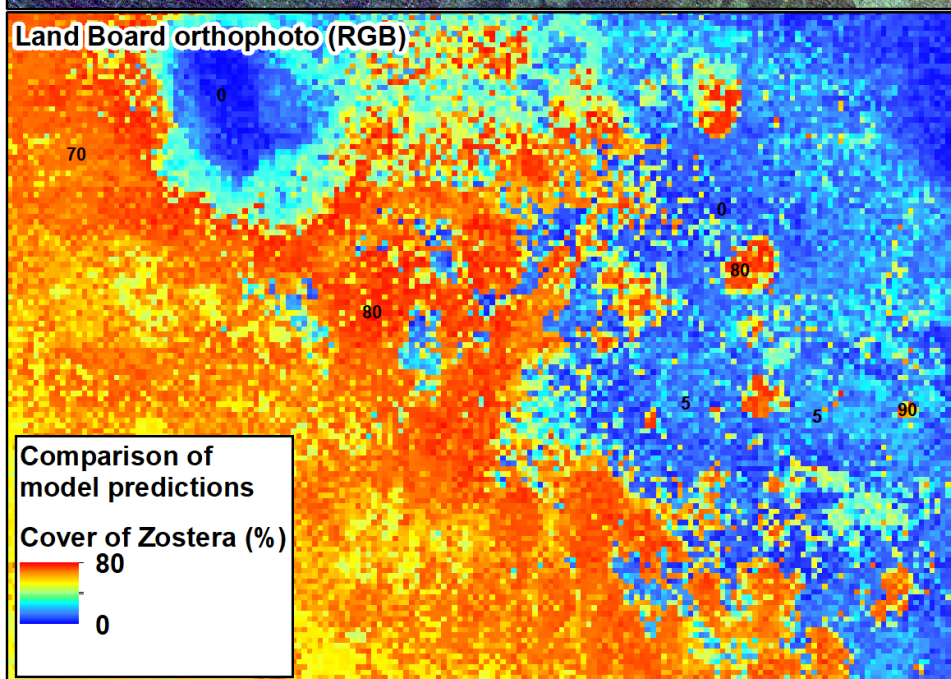
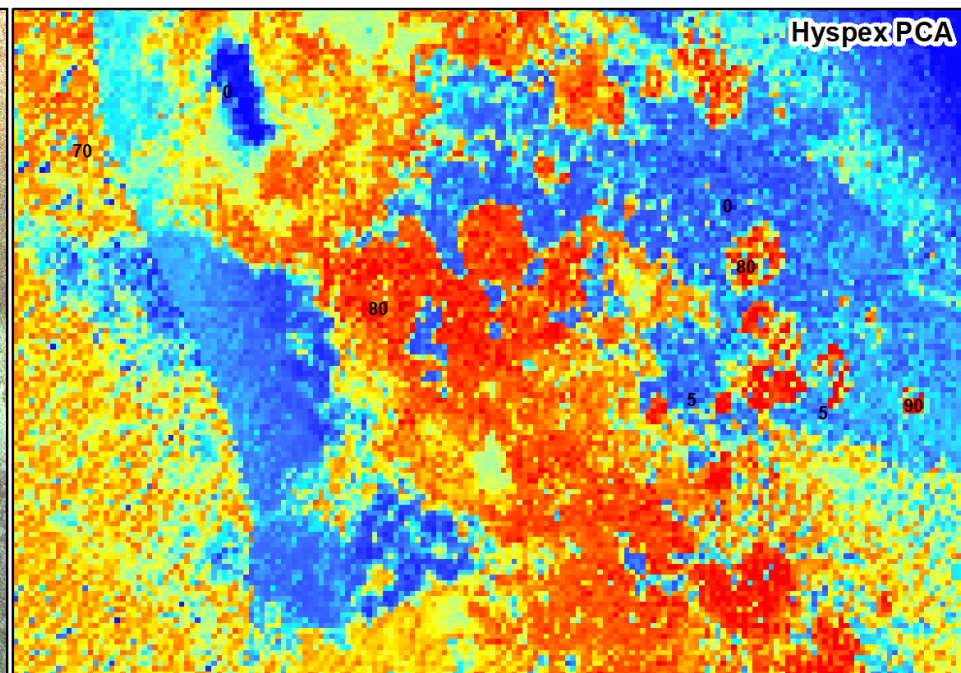
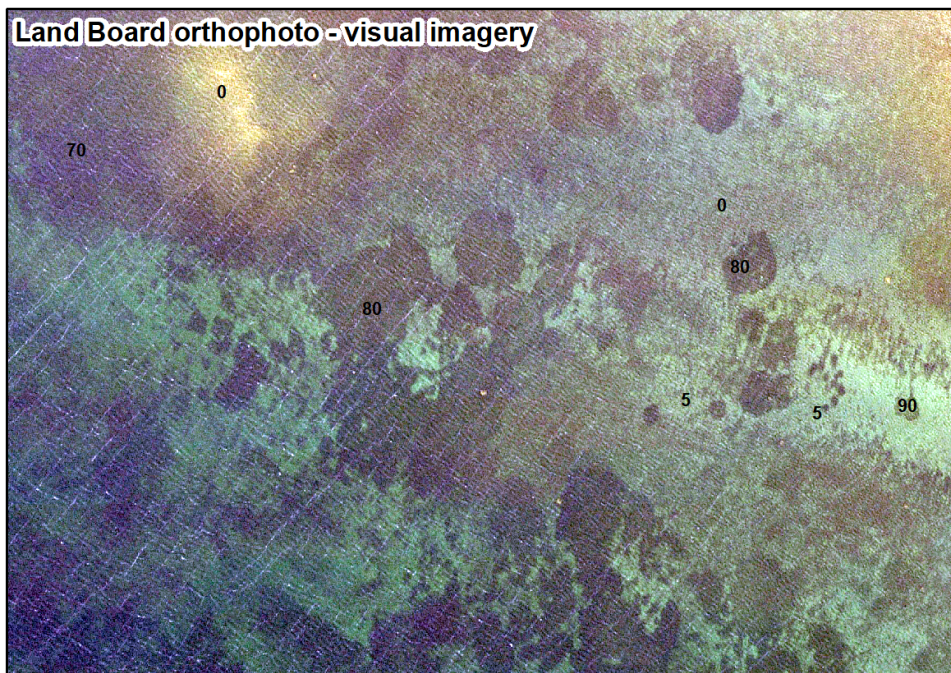
-  mixed
-  muddy
-  rock & boulder
-  sand









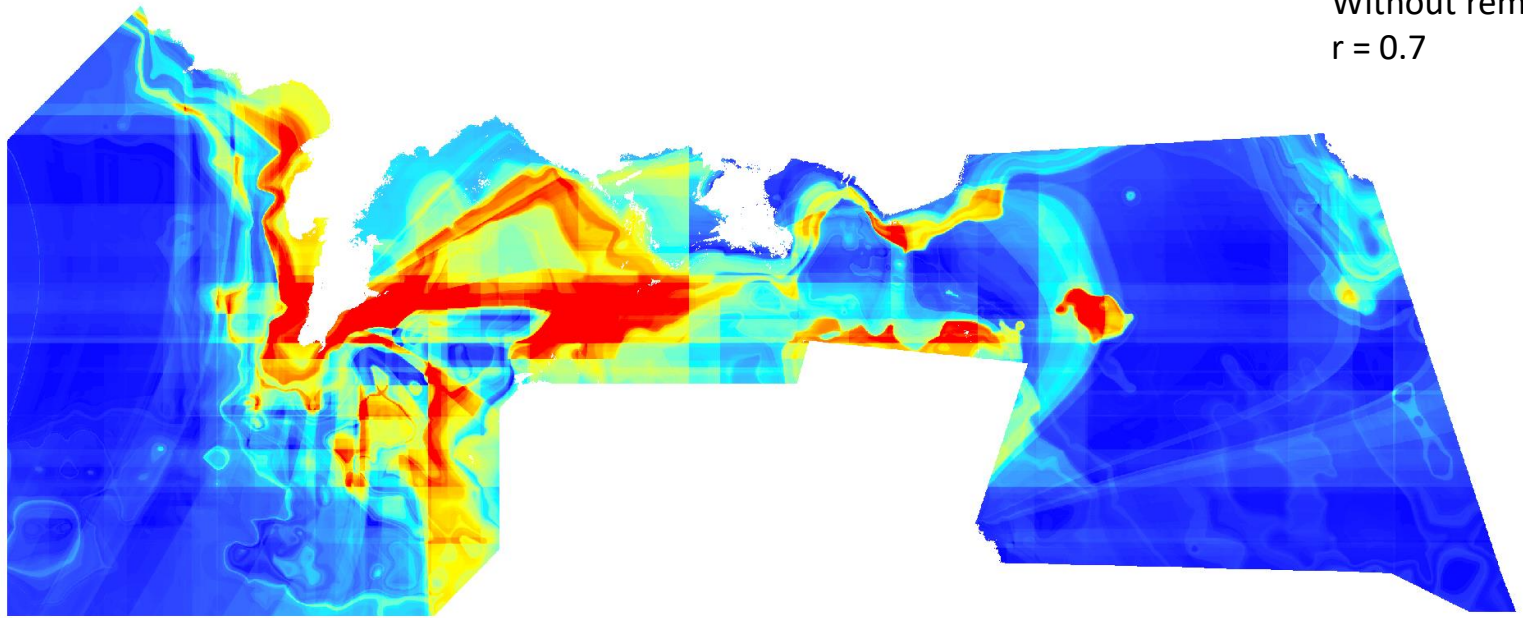




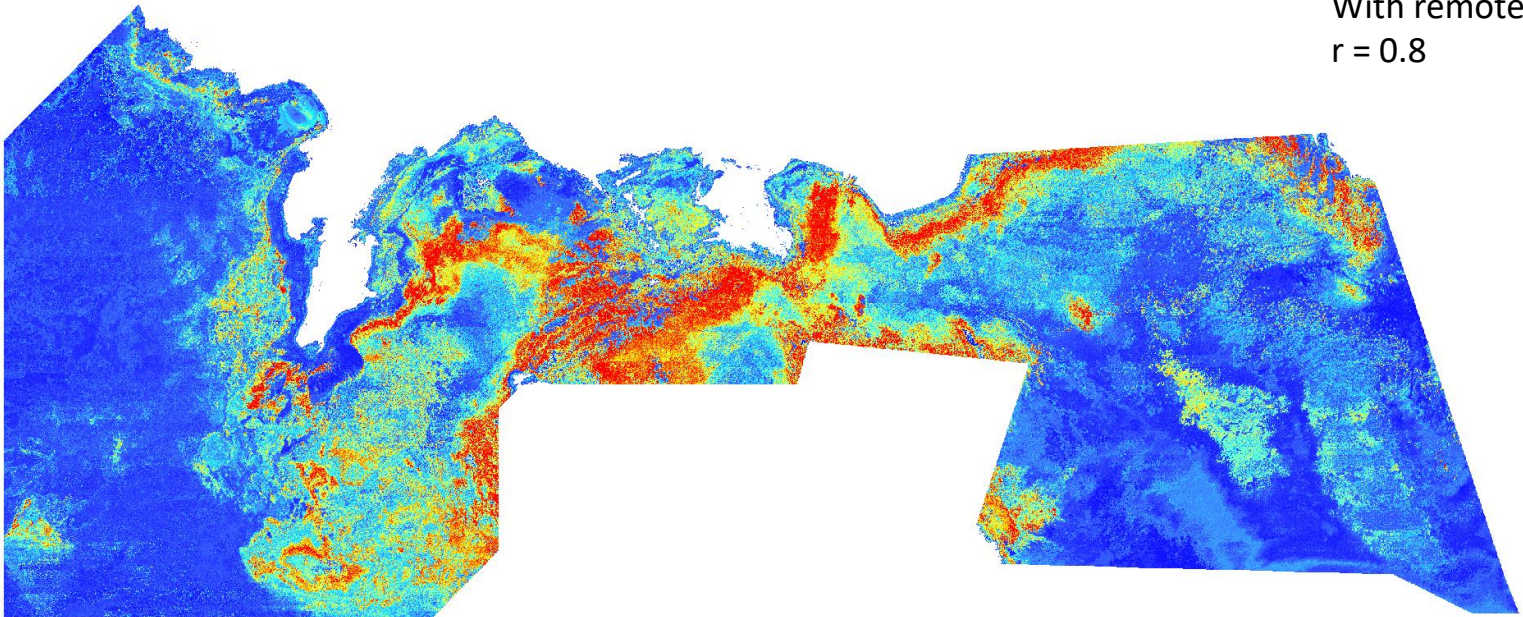
Setting/model		
Number of training cells per site	9	4
Area-based or pooled model	Area-based	Pooled model
Modeling algorithm	RF	BRT, NN
Orthophoto or Hypsax (2.3.1)	Hypsax	Orthophoto
Drone or Hypsax – visual features	Drone	Hypsax
Drone or Hypsax – modeling	?Hypsax?	?Drone?
Hypsax PCA12 or 12 bands	PCA12	12 bands

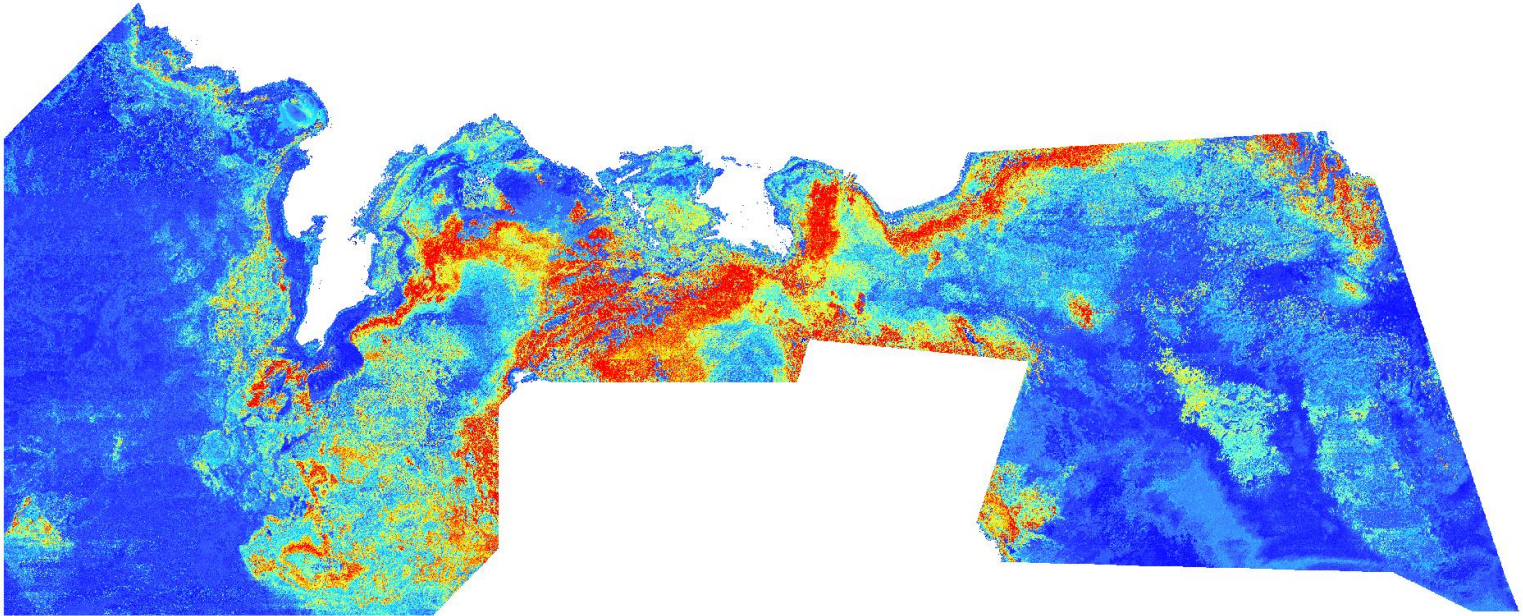
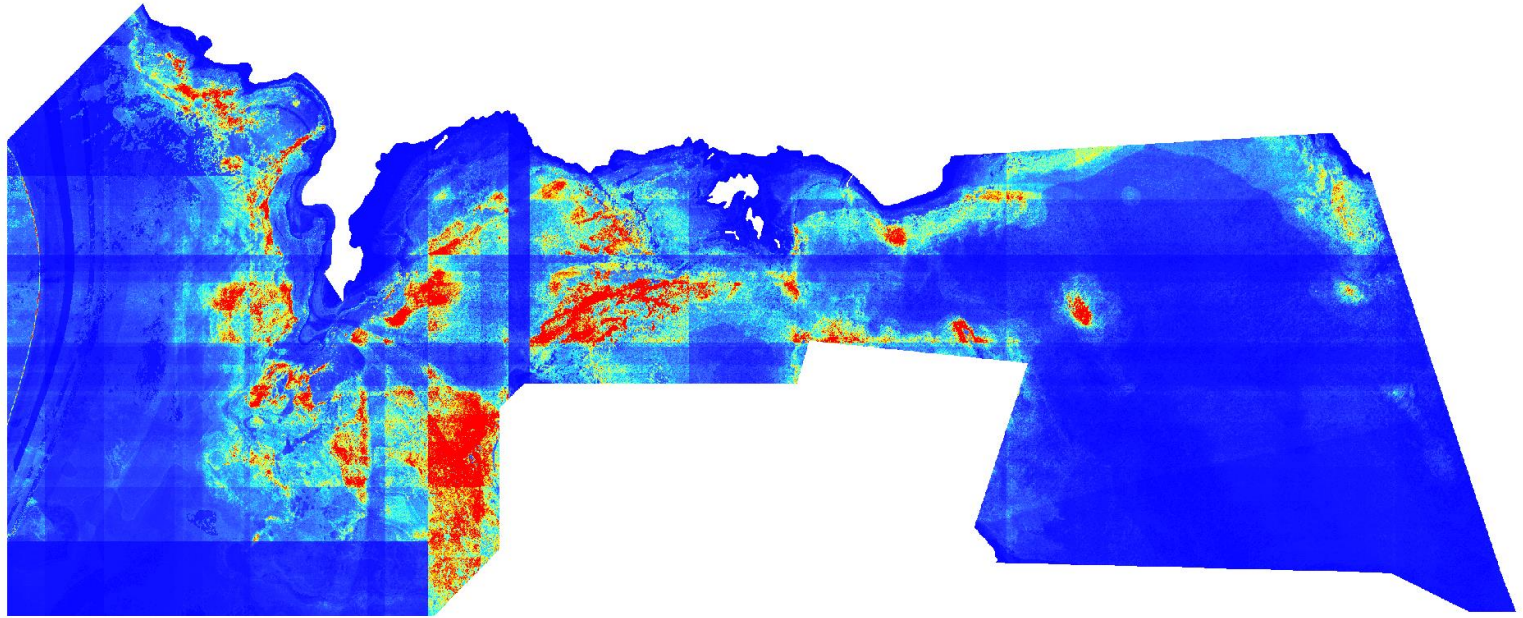
- Increase in prediction accuracy using remote sensing in 2.3.1
 - r : 0.7 \rightarrow 0.8
 - MAE: 19 \rightarrow 15
 - Class accuracy: 0.58 \rightarrow 0.64
- Mathematical validation may fail to reveal overfitting artifacts and other un-natural patterns

Without remote sensing
 $r = 0.7$



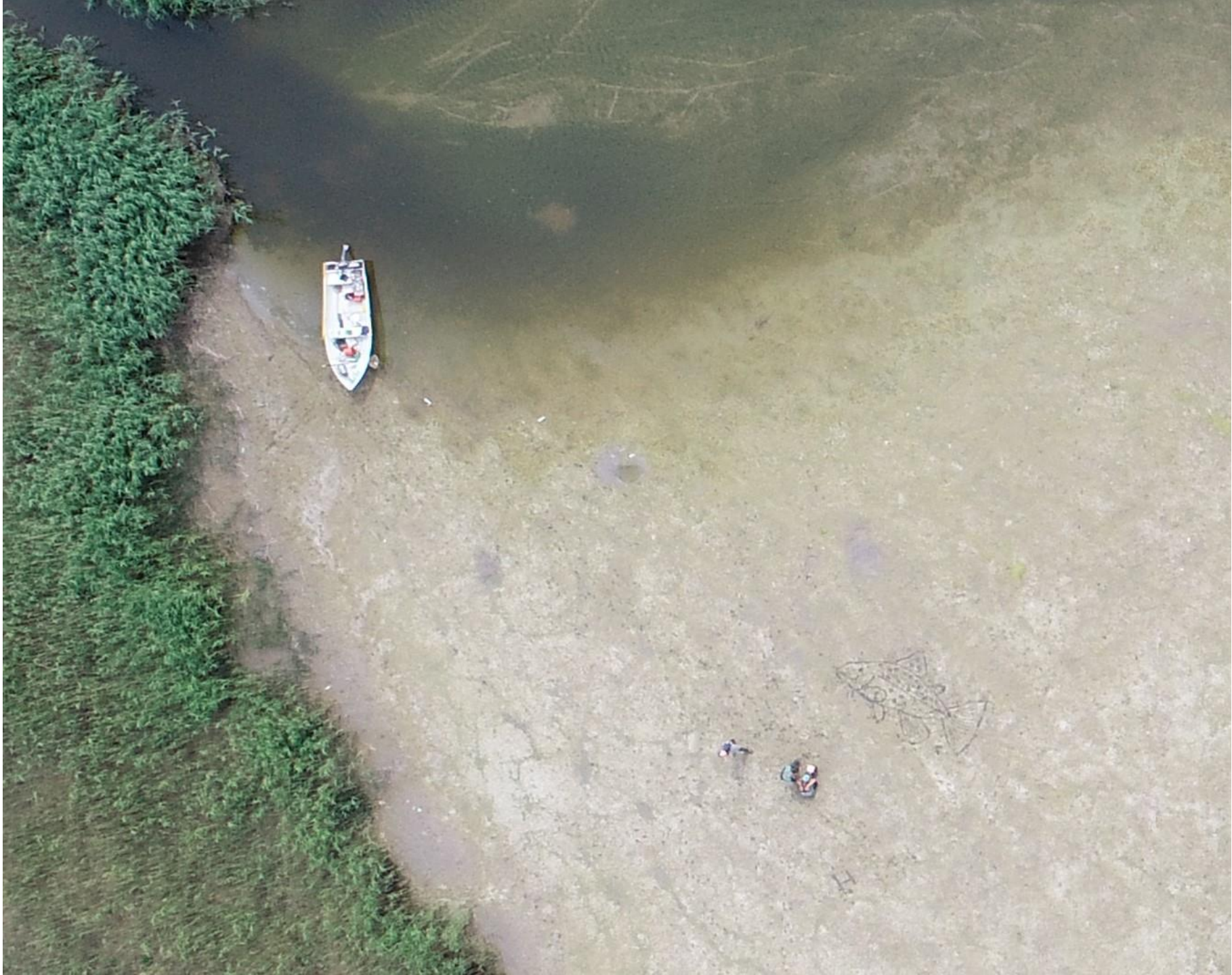
With remote sensing
 $r = 0.8$





Conclusions

- Remote sensing (RS) increases the accuracy of benthic distribution maps
- RS is the only way to produce maps of very sparsely and spatially irregularly sampled shallow water bodies
- Different RS products are usable depending on
 - Spatial scale of a study area
 - Aim of the study
 - Sizes of features to be mapped
 - Availability and quality of RS data
 - Available resources (time, money)



Uuringu tellis ja uuringut rahastab Eesti Teadusagentuur Euroopa Regionaalarengu Fondist toetatava programmi „Valdkondliku teadus- ja arendustegevuse tugevdamine“ (RITA) tegevuse 1 „Strateegilise TA tegevuse toetamine“ kaudu. Uuring valmib Keskkonnaministeeriumi eesmärkide elluviimiseks.

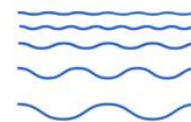
This study was financially supported by the European Regional Development Fund within National Programme for Addressing Socio-Economic Challenges through R&D (RITA).



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