

# ADRIENNE

## Mapping biota and predicting future changes

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# Mapping and modeling in ADRIENNE

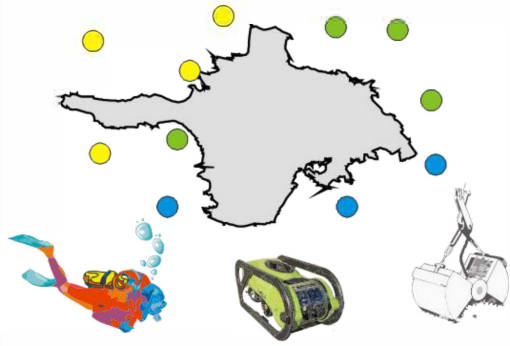
## Why spatial predictive modelling?

- ✦ Generate spatially continuous map layers
- ✦ Quantify changes in the distribution of species, habitats or ecosystem services due to climate change and human impact
- ✦ Generate spatially continuous input for the GIS Assessment Portal

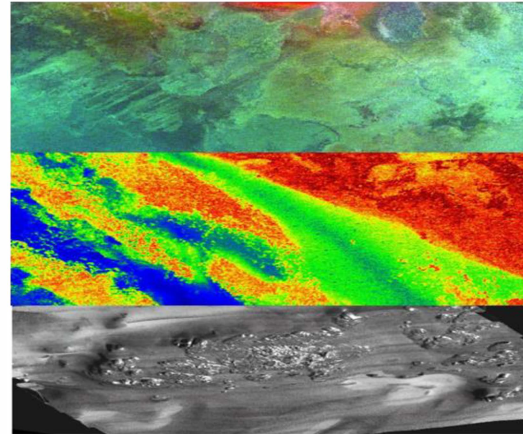
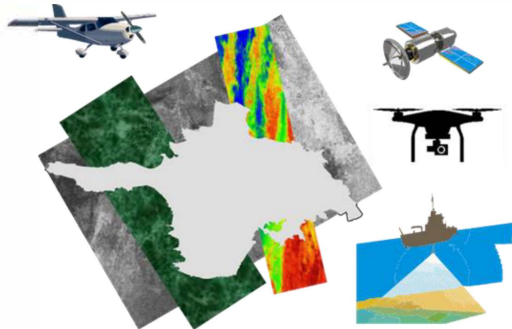
# ✦ Distribution modeling of macroscopic seabed plants and invertebrates



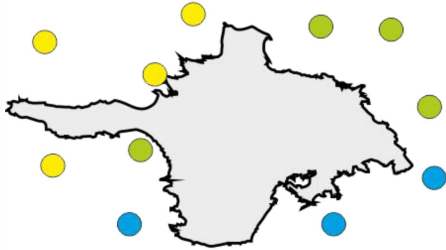
**On-site sampling: point-wise data of substrate, biota**



**Remote sensing, hydrodynamic modeling: continuous layers of physical variables**



**Response variable**  
point data of species



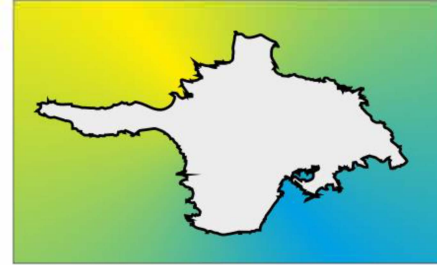
**Predictor variables**  
GIS layers of  
environmental data



**Model**  
GAM  
BRT  
RF

**Prediction**

occurrence of species

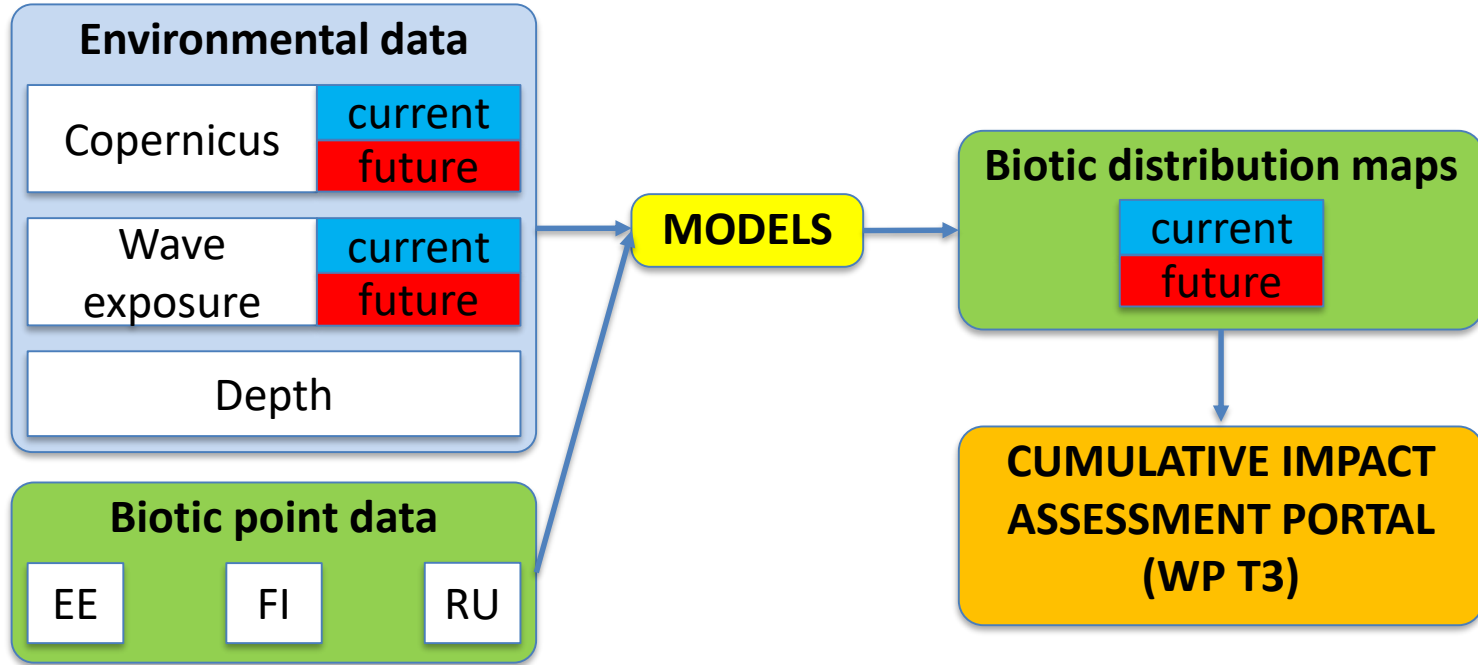


**Model assessment**

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



# Concept

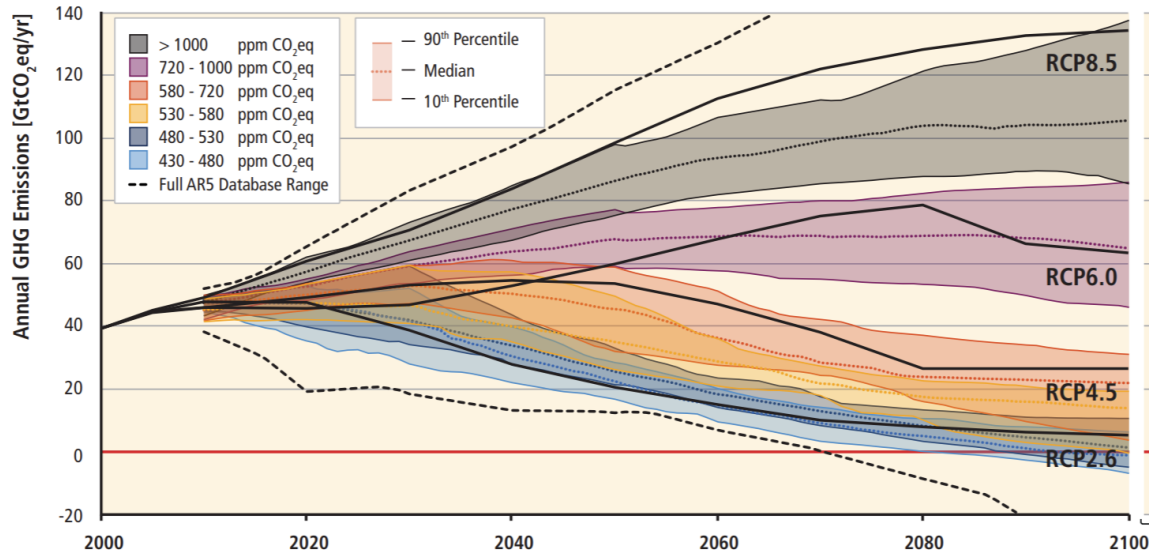


- What is the impact of human activities on biota today and in future?
- What groups of biota and where are positively/negatively affected by climate change?

# Scenarios of climate change and human pressure

- ✦ Climate scenario IPCC RCP6.0
- ✦ Warmer winters in the northern Europe, increased storminess

GHG Emission Pathways 2000-2100: All AR5 Scenarios



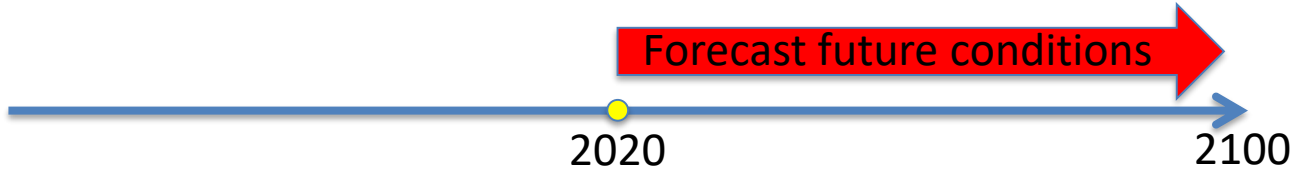
IPCC, 2014: Summary for Policymakers

**ipcc**  
INTERGOVERNMENTAL PANEL ON  
climate change

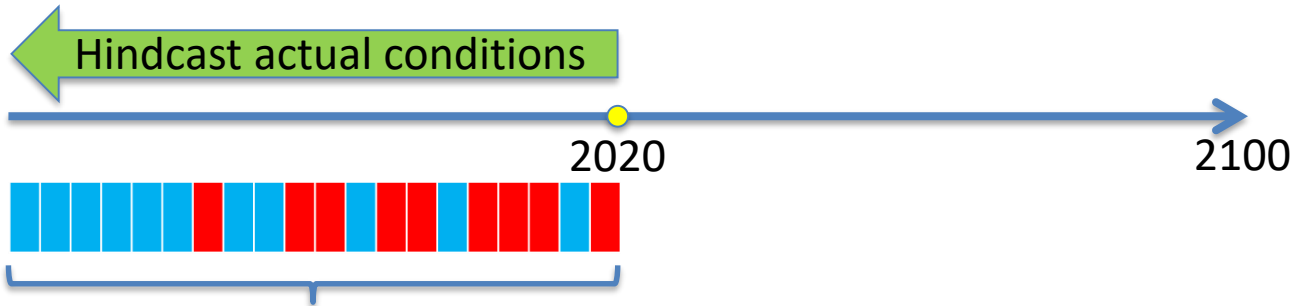


## 2 approaches to model future distribution of species:

- ✦ Forecast environmental variables based on future climate



- ✦ Hindcast environmental variables based on real climate data → select the years of „current“ and „future“ climate based on resemblance



Select data from years resembling „**current**“ or „**future**“ conditions



## ✦ We use the „hindcast“ approach

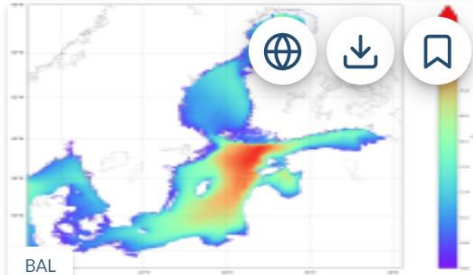
- Hindcast data is based on real measured atmospheric and marine data → adequate realization of interplay of different parameters (e.g. salinity, temperature, oxygen, chlorophyll, nutrients)
- There have been many years similar to what is projected in IPCC RCP6.0 in the recent 30 years (i.e. warm, moist winters)



# ✦ Copernicus hindcast data: salinity, temperature, oxygen, ice, current velocity, nutrients, chlorophyll, Secchi depth, wave height



Copernicus Marine Service



## Baltic Sea Wave Hindcast

BALTICSEA\_REANALYSIS\_WAV\_003\_015

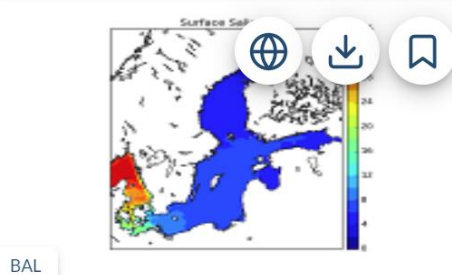
SWH MWT VMDR VSDXY WW SW1 SW2 ⓘ

From To  
1993-01-01 2020-06-30

2 km x 2 km

Model assimilation  
None

Surface only  
ⓘ hourly instantaneous  
Sub-setting WMS



## Baltic Sea Physics Reanalysis

BALTICSEA\_REANALYSIS\_PHY\_003\_011

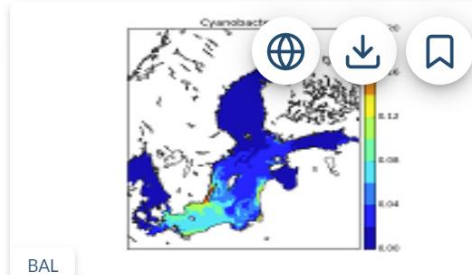
T bottomT S SSH UV MLD SIC SIT ⓘ

From To  
1993-01-01 2019-12-31

4 km x 4 km

Model assimilation

56 depths level  
ⓘ hourly instantaneous - daily mean - monthly mean  
Sub-setting WMS



## Baltic Sea Biogeochemistry Reanalysis

BALTICSEA\_REANALYSIS\_BIO\_003\_012

CHL O2 NO3 PO4 NH4 ⓘ

From To  
1993-01-01 2019-12-31

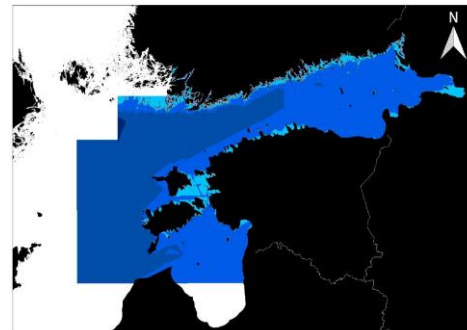
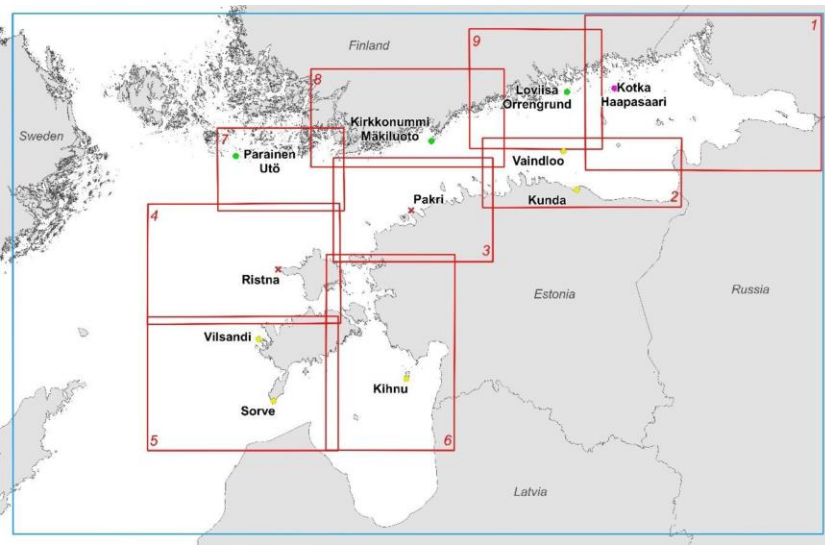
4 km x 4 km

Model assimilation

56 depths level  
ⓘ daily mean - monthly mean  
Sub-setting WMS

# ✦ Simplified wave model for current and future climate

- Modeling by AquaBiota Water Research (Sweden)
- Based on Estonian and Finnish wind data



EstExp, mosaic of all grids; future scenario  
0 50 100 200 Kilometers



## ✦ Human pressure scenario - eutrophication

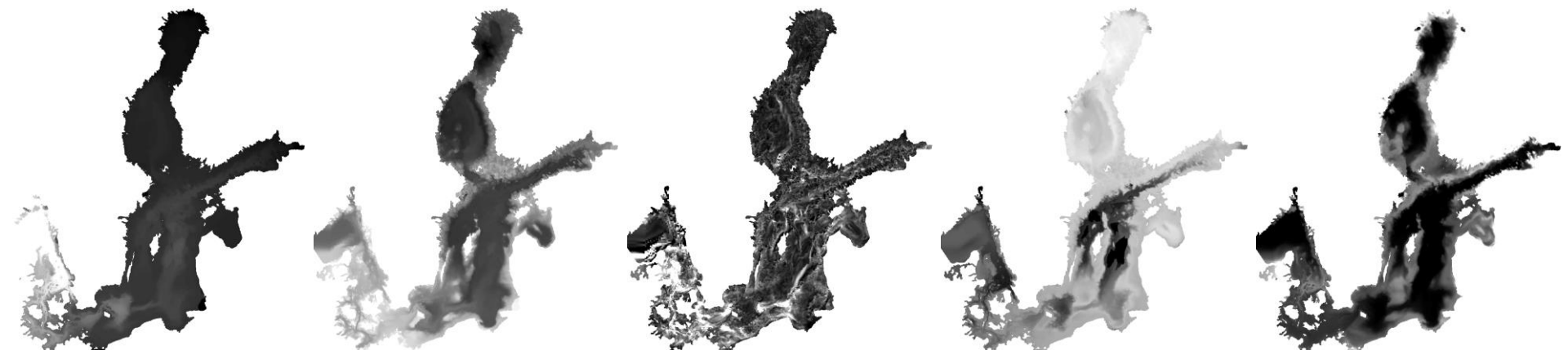
- Current
- Eutrophication mitigation: 25% reduction of nutrient concentrations (HELCOM BSAP)

## ✦ Full set of scenarios

- Current
- Climate change
- Eutrophication mitigation
- Climate change + eutrophication mitigation

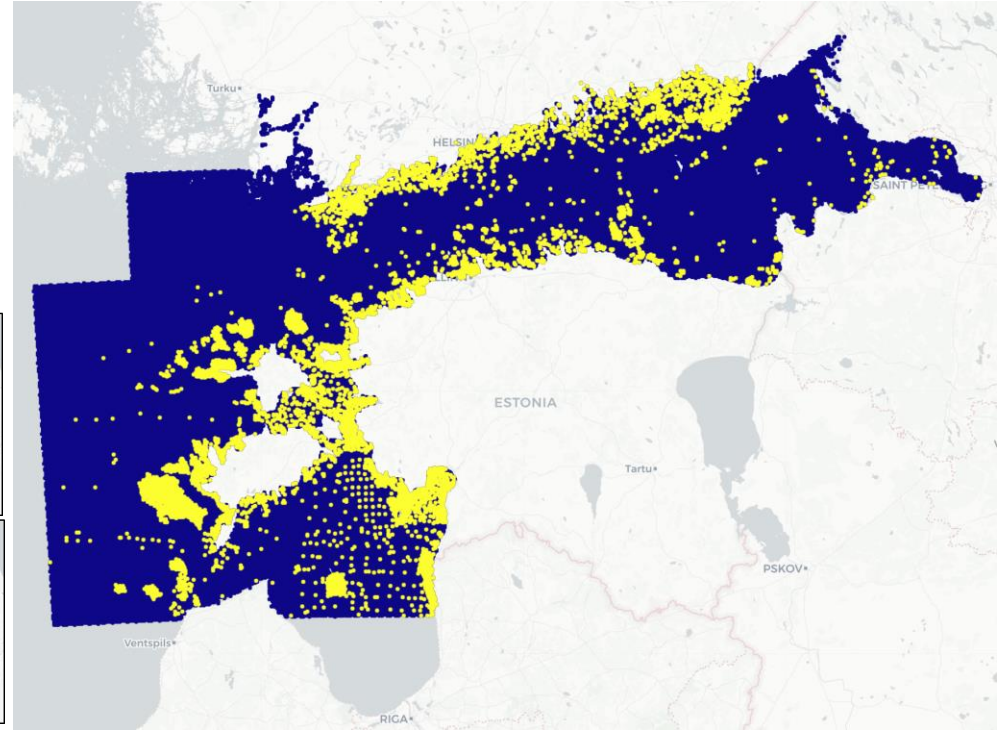
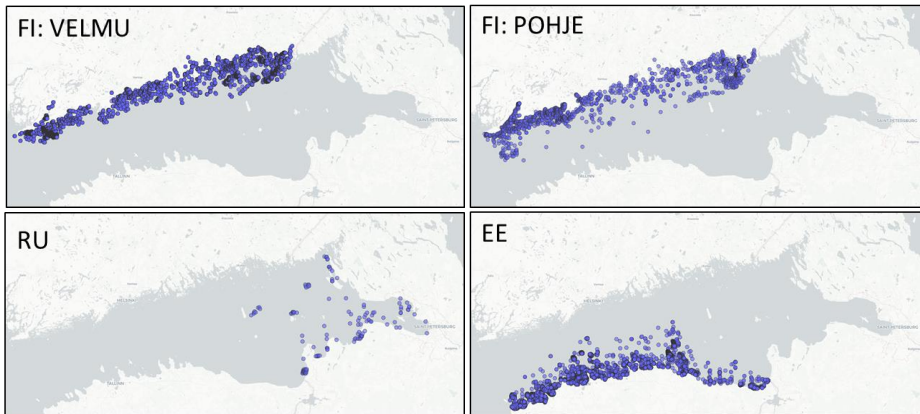
# Environmental data

- ✦ Final selection of variables: depth, wave exposure, salinity, temperature, Secchi depth, wave height, nitrates, phosphates, ice cover, chlorophyll
- ✦ All data aggregated to 1 km grid
- ✦ Values for climate change (all except depth) and eutrophication mitigation (nitrates, phosphates) scenarios



# Benthos data

- ✦ Data from FI, RU, EE, LV sources
- ✦ All data aggregated to 1 km grid, full grid 77495 cells
- ✦ 6770 cells with benthos data



## Selection of species/groups for modeling

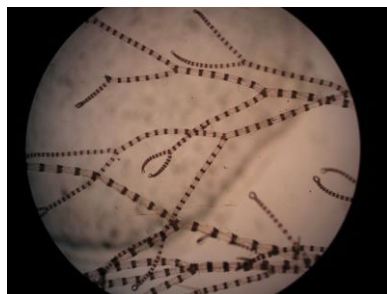
- ✦ Groups selected based on occurrence rate and ecological relevance
- ✦ 3 levels of groups
  - Group 1 (n = 42): lowest level, mainly species or genus  
*Amphibalanus improvisus, Ampullaceana balthica, Battersia arctica, Ceramium, Cerastoderma, Ceratophyllum demersum, charophytes, Chironomidae, Chorda filum, Cladophora glomerata, Cladophora rupestris, Coccotylus truncatus, Dictyosiphon foeniculaceus, Dreissena polymorpha, Fucus, Furcellaria lumbricalis, Gammarus, Halicryptus spinulosus, Hediste diversicolor, Hildenbrandia, Idotea, Jaera, Limecola balthica, Marenzelleria, Monoporeia affinis, Mya arenaria, Myriophyllum, Mytilus trossulus, Najas marina, Oligochaeta, Potamogeton perfoliatus, PylaiellaEctocarpus, Rhodomela confervoides, Ruppia, Saduria entomon, Stictyosiphon tortilis, Stuckenia, Theodoxus, Ulva, Vertebrata fucoides, Zannichellia, Zostera marina*
  - Group 2 (n = 14): class, life form, freshwater groups  
drifting macrophytes, epifaunal bivalves fresh, filamentous brown algae, filamentous green algae, filamentous red algae, hydrozoa, infaunal bivalves, infaunal bivalves fresh, other *Polychaeta*, snails, snails fresh, thick brown algae, thick red algae, vascular plants
  - Group 3 (n = 1): filamentous algae



*Fucus vesiculosus*



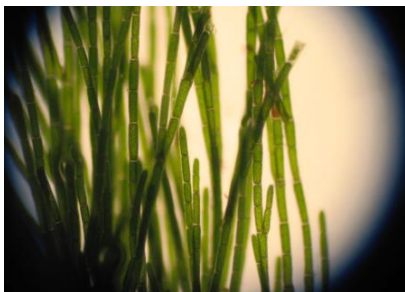
*Furcellaria lumbricalis*



*Ceramium spp*



*Ulva spp*



*Cladophora spp*



*Vertebrata fucoides*



*Pylaiella/Ectocarpus*



*Chorda filum*



*Mytilus trossulus*



*Dreissena polymorpha*



*Amphibalanus improvisus*

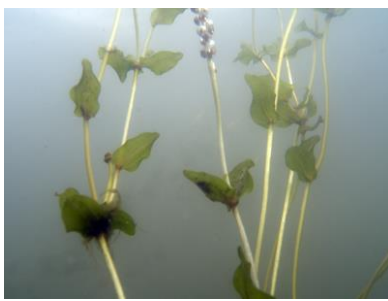


*Gammarus spp*

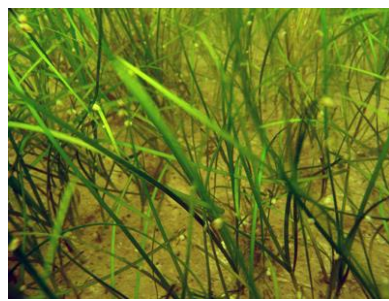




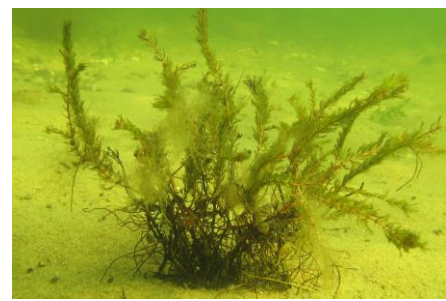
charophytes



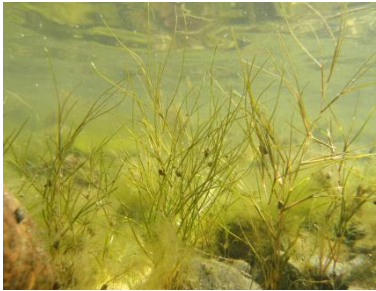
*Potamogeton perfoliatus*



*Zostera marina*



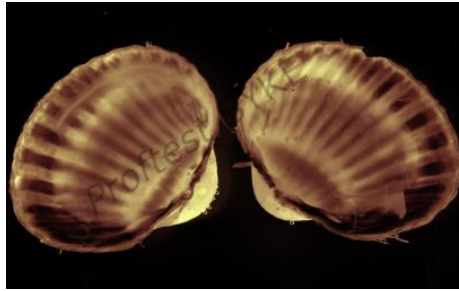
*Myriophyllum spicatum*



*Stuckenia pectinata*



*Ruppia maritima*



*Cerastoderma glaucum*



*Mya arenaria*



*Limecola balthica*



*Heiste diversicolor*

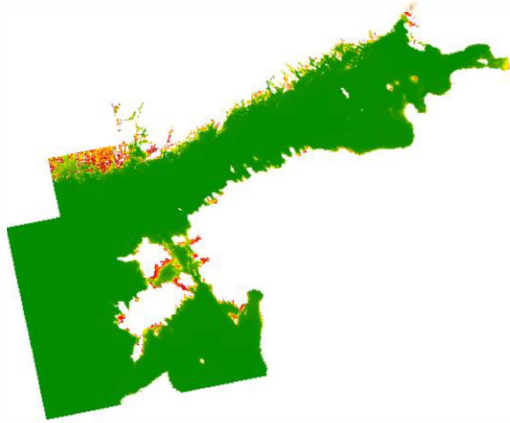


Chironomidae

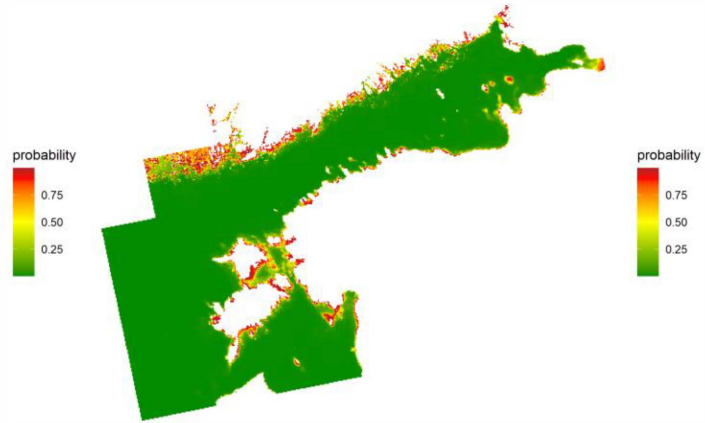
# Distribution modeling

- ✦ 57 species/groups
- ✦ Boosted regression trees (BRT) modeling algorithm
- ✦ Models trained on „current“ dataset
- ✦ 4 spatial predictions for each species/groups
  - Current
  - Climate change **DRAFT**
  - Eutrophication mitigation **DRAFT**
  - Climate change + eutrophication mitigation **DRAFT**
- ✦ Output of model predictions: probability of occurrence (0...1)
- ✦ Probability of occurrence converted to binary presence-absence

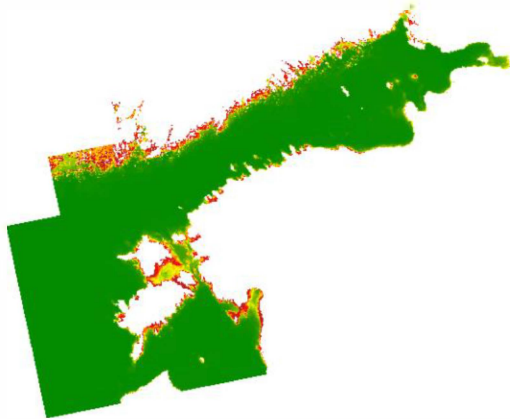
charophytes  
current | mean prob. 0.052



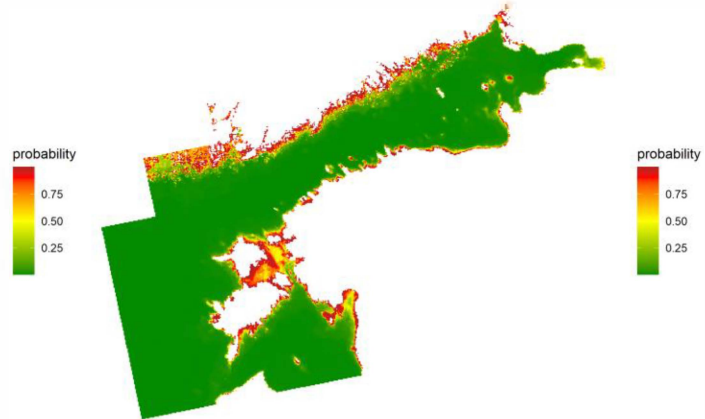
charophytes  
eutrophication mitigation | mean prob. 0.083



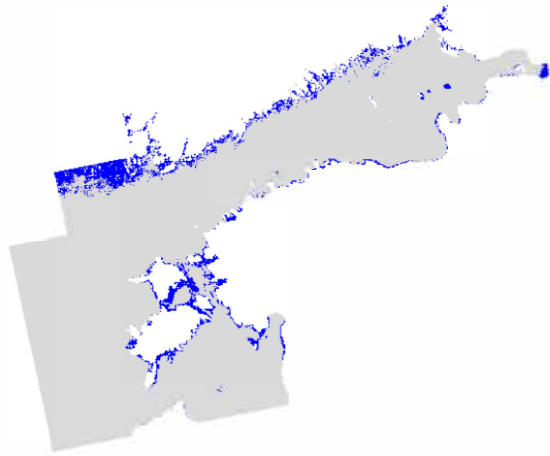
charophytes  
climate change | mean prob. 0.083



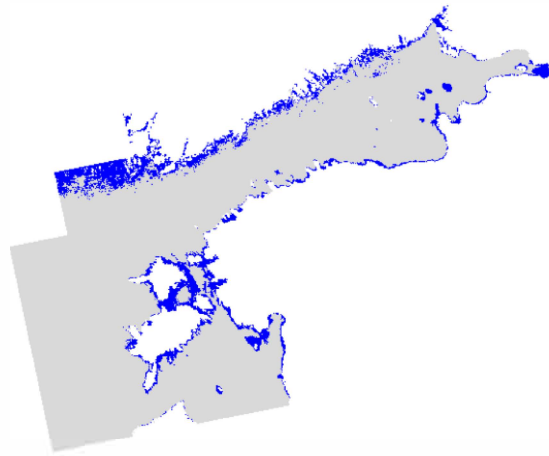
charophytes  
climate change + eutrophication mitigation | mean prob. 0.117



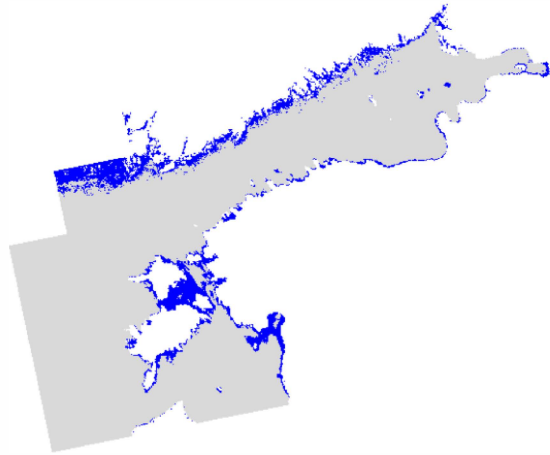
charophytes  
current | 5310 km2



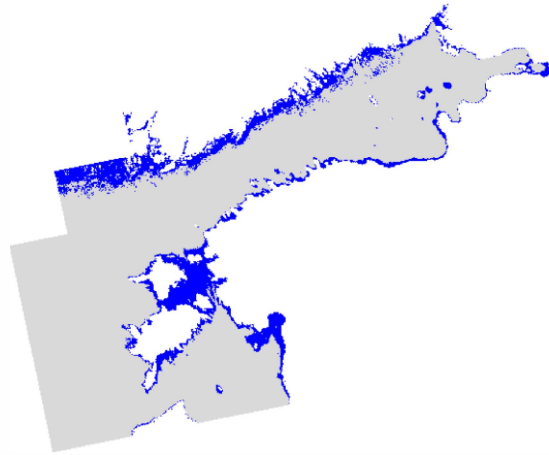
charophytes  
eutrophication mitigation | 8291 km2



charophytes  
climate change | 8240 km2



charophytes  
climate change + eutrophication mitigation | 11113 km2



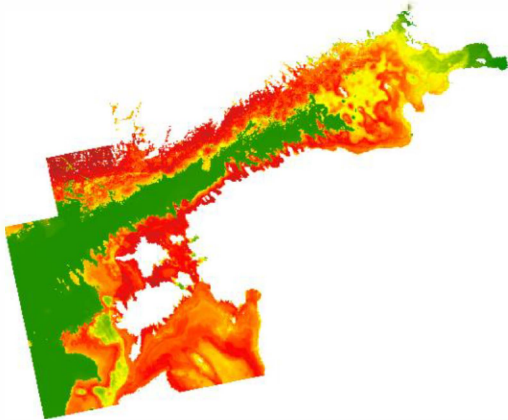
limecola\_balthica  
current | mean prob. 0.493



limecola\_balthica  
eutrophication mitigation | mean prob. 0.504



limecola\_balthica  
climate change | mean prob. 0.524



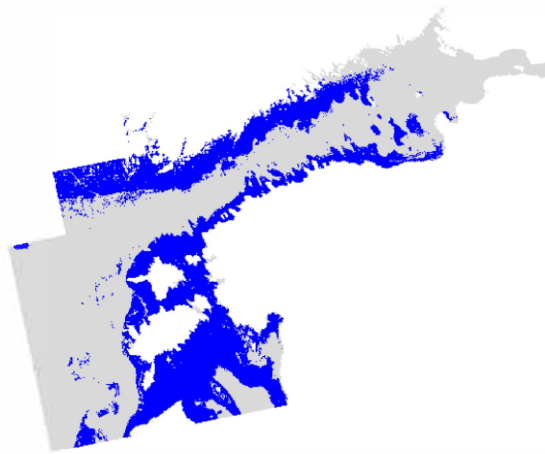
limecola\_balthica  
climate change + eutrophication mitigation | mean prob. 0.535



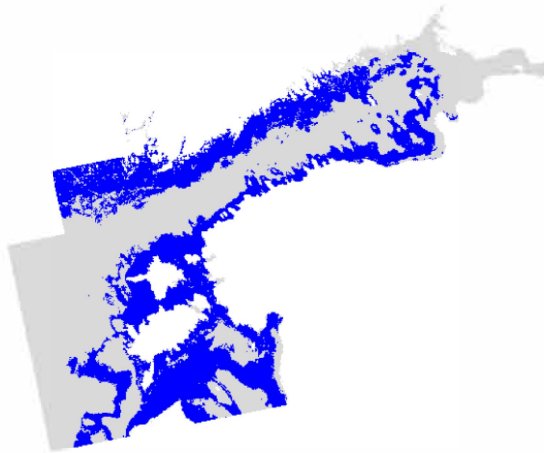
limecola\_balthica  
current | 28750 km<sup>2</sup>



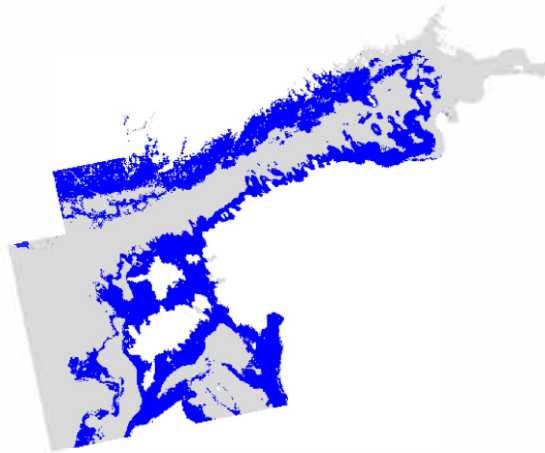
limecola\_balthica  
eutrophication mitigation | 27759 km<sup>2</sup>

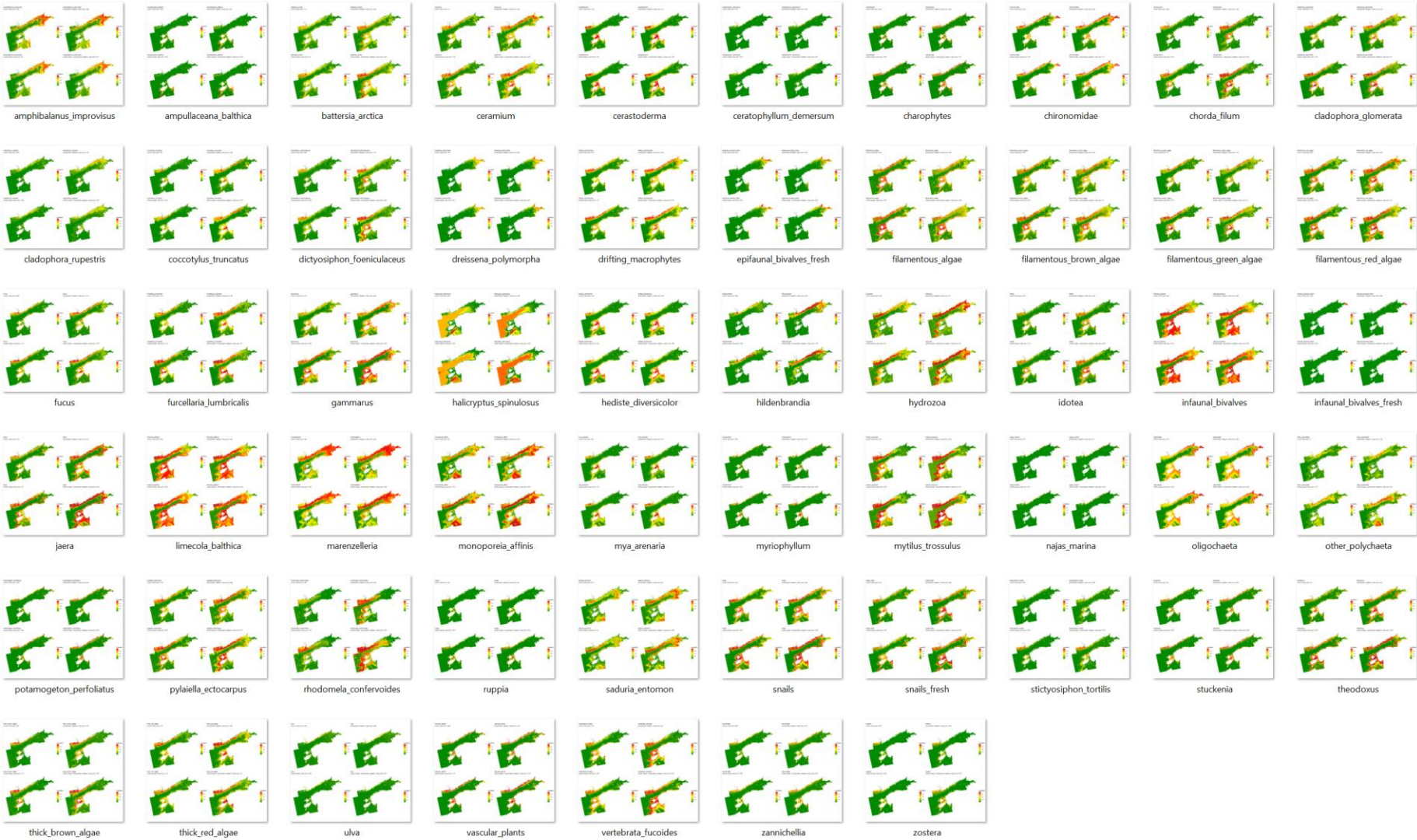


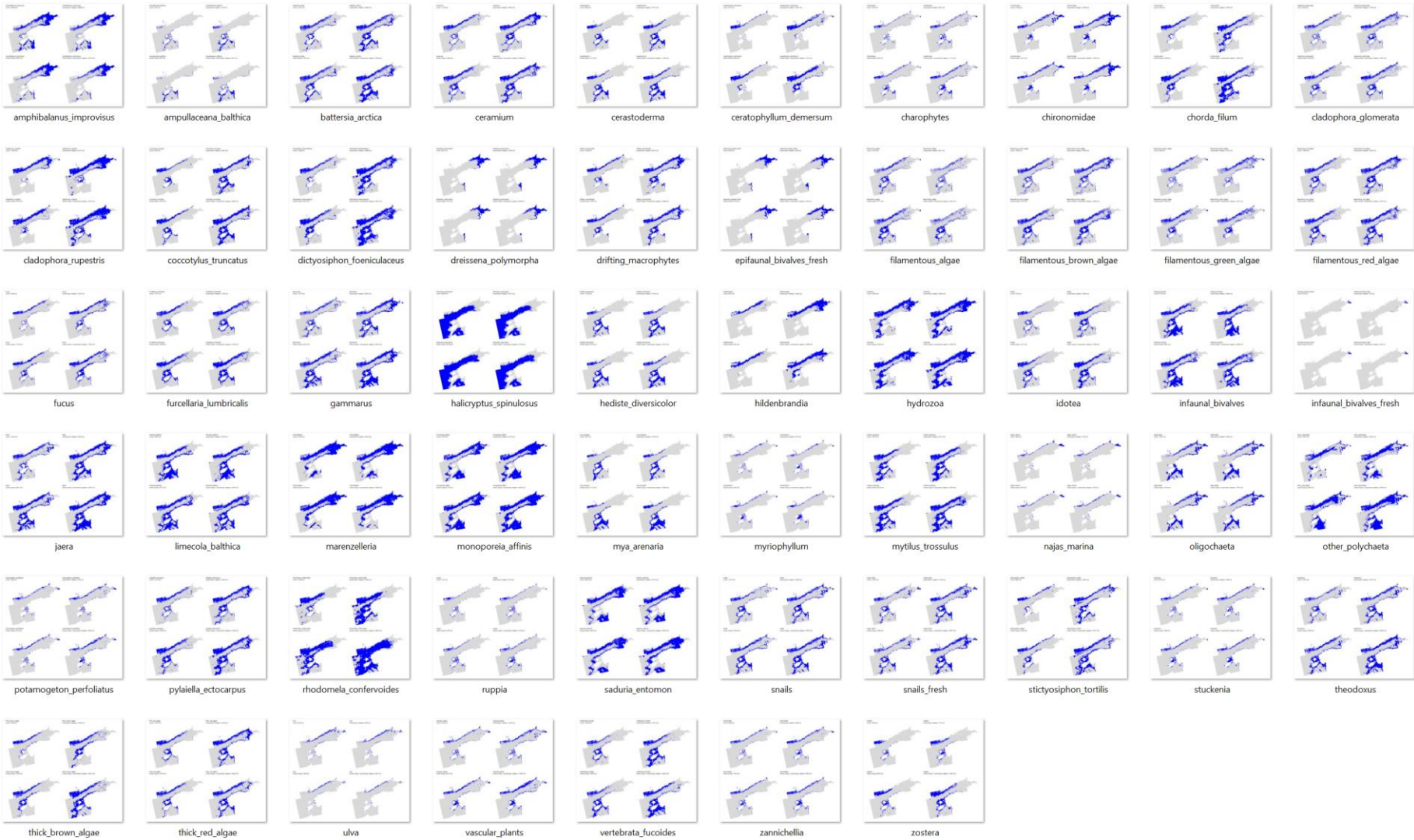
limecola\_balthica  
climate change | 30777 km<sup>2</sup>



limecola\_balthica  
climate change + eutrophication mitigation | 30579 km<sup>2</sup>









# THANK YOU!

<https://adrienne.ut.ee/>

