

BioSpace25 - Biodiversity insight from Space
10 - 14 February 2025 | ESA-ESRIN | Frascati - Italy

Phytoplankton assemblage structure off southwestern Iberia: Combining complementary approaches to assess variability and underlying drivers/predictors

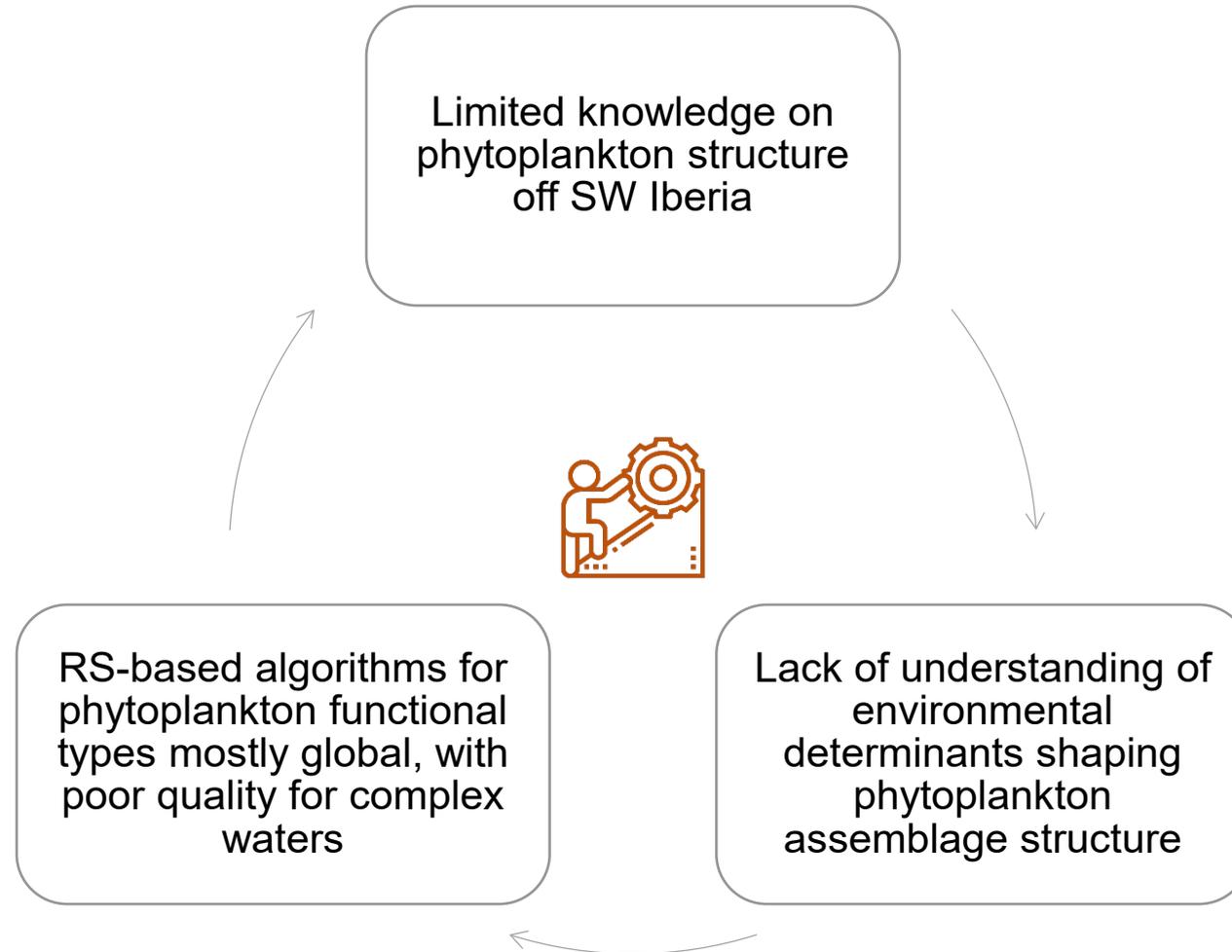
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Why study phytoplankton off SW Iberia?

- Key primary producers, regulators of ecosystem functioning and services, and relevant indicators of environmental status
- Different species-specific functional traits and niche preferences lead to variable responses to environmental forcing
- Region characterized by high levels of mesoscale variability and vulnerability to climate change (Soares et al., 2023)

What are the current challenges?





Study objectives

1. Assess spatial-temporal variability patterns of phytoplankton assemblages off SW Iberia
2. Identify underlying environmental drivers and predictors
3. Evaluate the performance of satellite algorithms used to derive phytoplankton composition at a regional scale

Methods: field sampling



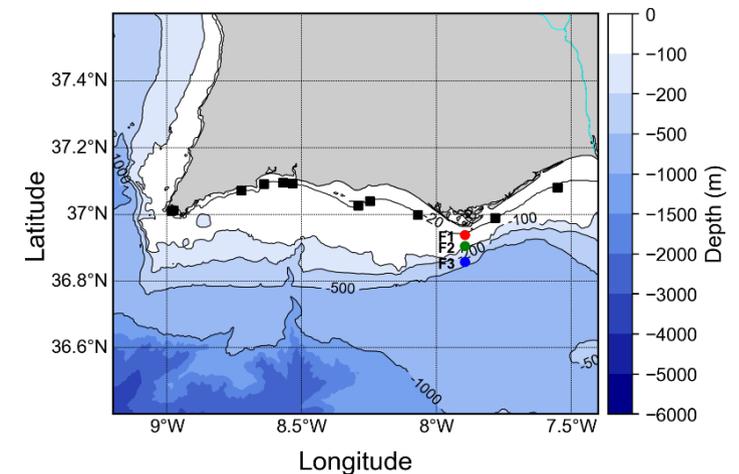
Regular sampling

- 3 stations (F1, F2 and F3)
- Coastal-offshore transect
- July 2012 – July 2014 (10 sampling dates)
- 3 depth levels: surface, 0.3*Secchi depth, 0.6*Secchi depth
- Sampling time synchronous with MODISA passage ($\pm 3h$)

MarBIS oceanographic cruise

- 12 stations (black squares)
- Alongshore transect
- 20 June – 11 July 2013
- 3 depth levels

Satellite algorithm validation



Methods: variables, data sources and analysis

 *In situ*  Remote sensing  Models

Physico-chemical: T, S, SDD, PAR, DO, NH_4^+ , NO_3^- , NO_2^- , PO_4^{3-} , SiO_4^{4-} , SPM

Phytoplankton-related variables:
Chl-a concentration, specific photosynthetic diagnostic pigments, abundance, biomass, community size structure and species composition

SST, PAR, Kd490, wind components and magnitude, Chl-a

NH_4^+ , NO_3^- , PO_4^{3-} , SiO_4^{4-} , MLD, geostrophic velocity

↪ Multivariate analysis (NMDS, PERMANOVA, CCA)

Methods: phytoplankton assemblage composition

- Contribution of size classes and functional types estimated from:
 - Abundance and biomass (epifluorescence and inverted microscopy)
 - Specific diagnostic pigment analysis (Hirata et al., 2011; Brewin et al., 2015 with weights tuned to the North Atlantic)
 - CHEMical TAXonomy analysis (CHEMTAX; Hayward et al., 2023)
- Comparison with satellite-based algorithms (3x3 matchup grid box; Xi et al., 2021), including the re-tuned three-component model by Brewin *et al.* (2010) with SST (Brewin et al., 2017), using type II linear regression

Methods: RS-derived phytoplankton composition



Daily data, 2012-2014

PSCs & PFTs



North Atlantic Ocean Colour Plankton, Reflectance, Transparency and Optics MY L3 daily observations

Home > Marine Data Store > Product

OCEANCOLOUR_ATL_BGC_L3_MY_009_113 (L3, 1x1 km)

Global Ocean Colour (Copernicus-GlobColour), Bio-Geo-Chemical, L3 (daily) from Satellite Observations (1997-ongoing)

Home > Marine Data Store > Product

OCEANCOLOUR_GLO_BGC_L3_MY_009_103 (L3, 4x4 km)

PSCs

Global Ocean Colour Plankton and Reflectances MY L3 daily observations

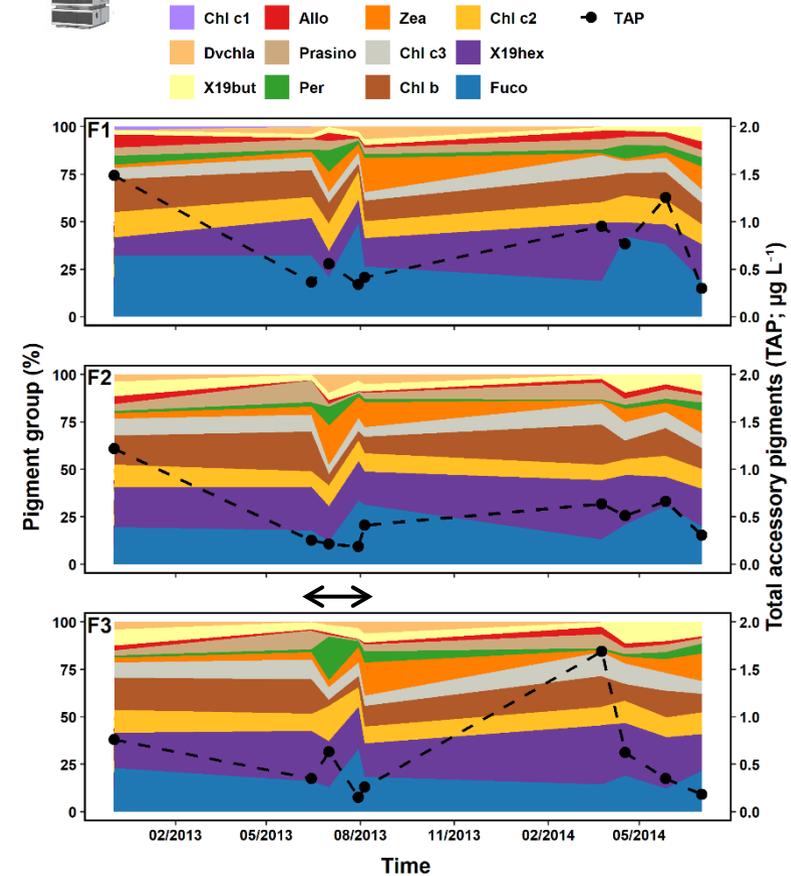
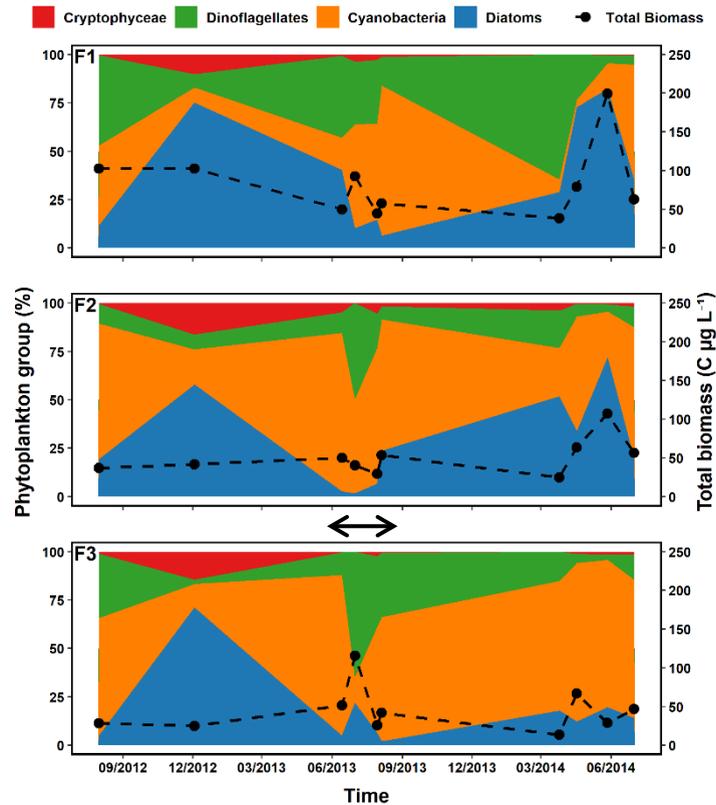
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OCEANCOLOUR_GLO_BGC_L3_MY_009_107 (L3, 4x4 km)



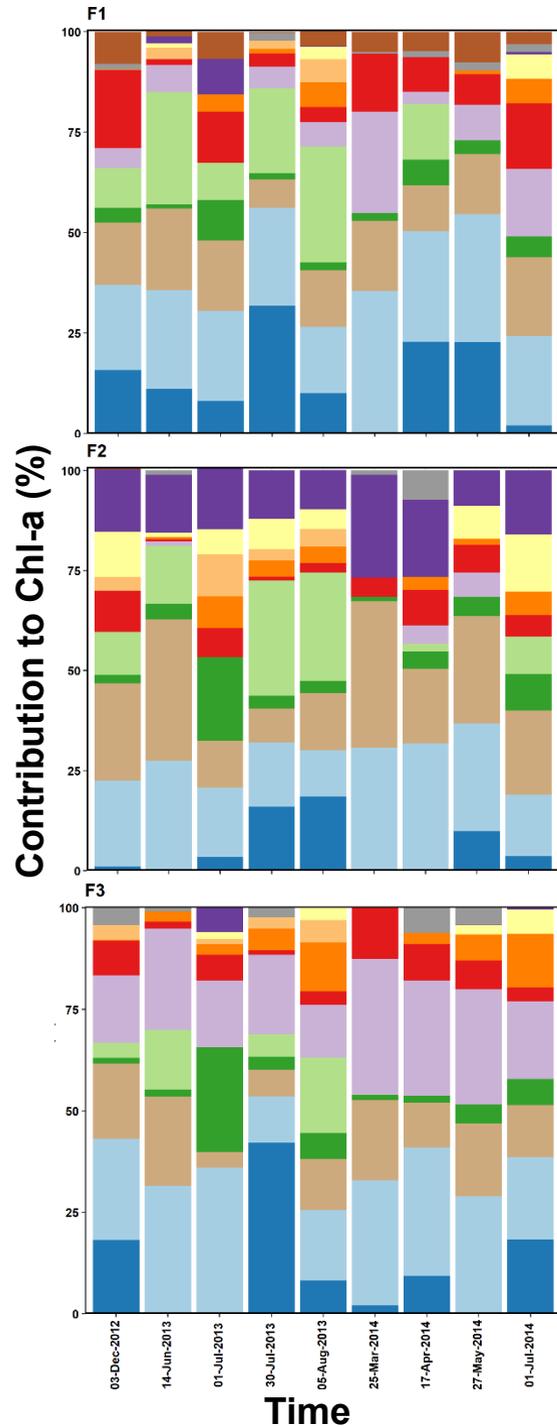
Processing frameworks and providers
Spatial extent and resolution
Available variables

Phytoplankton assemblage structure and pigment signatures



Chl: Chlorophyll, Dvchla: divinyl chlorophyll a; X19but: 19'-but-fucoxanthin; Allo: Alloxanthin; Prasino: Prasinoloxanthin; Per: Peridinin; Zea: Zeaxanthin; X19hex: 19'-hex-fucoxanthin; Fuco: Fucoxanthin

Phytoplankton assemblage structure from pigment signatures



Phytoplankton classes

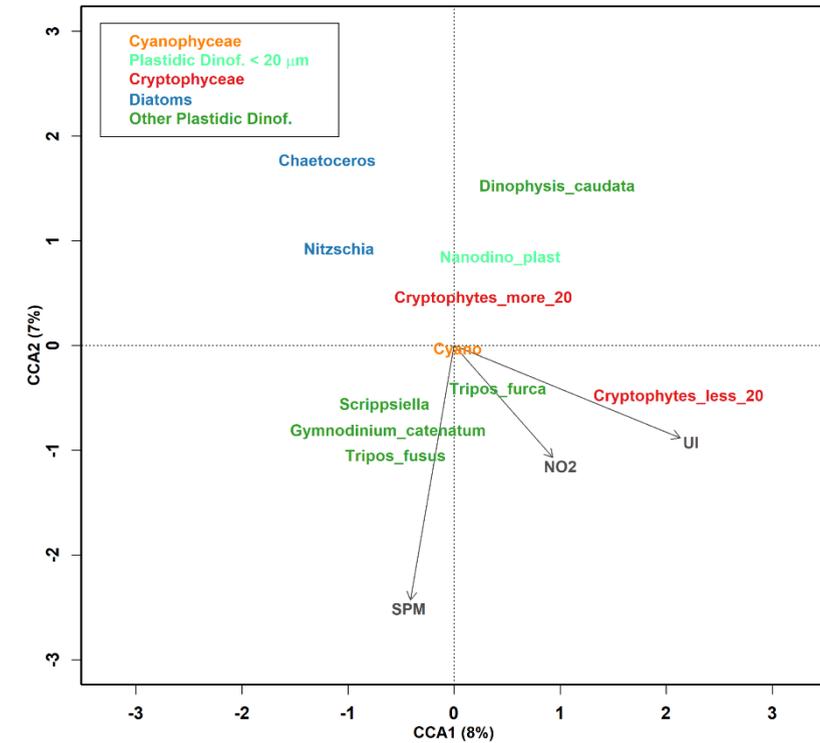
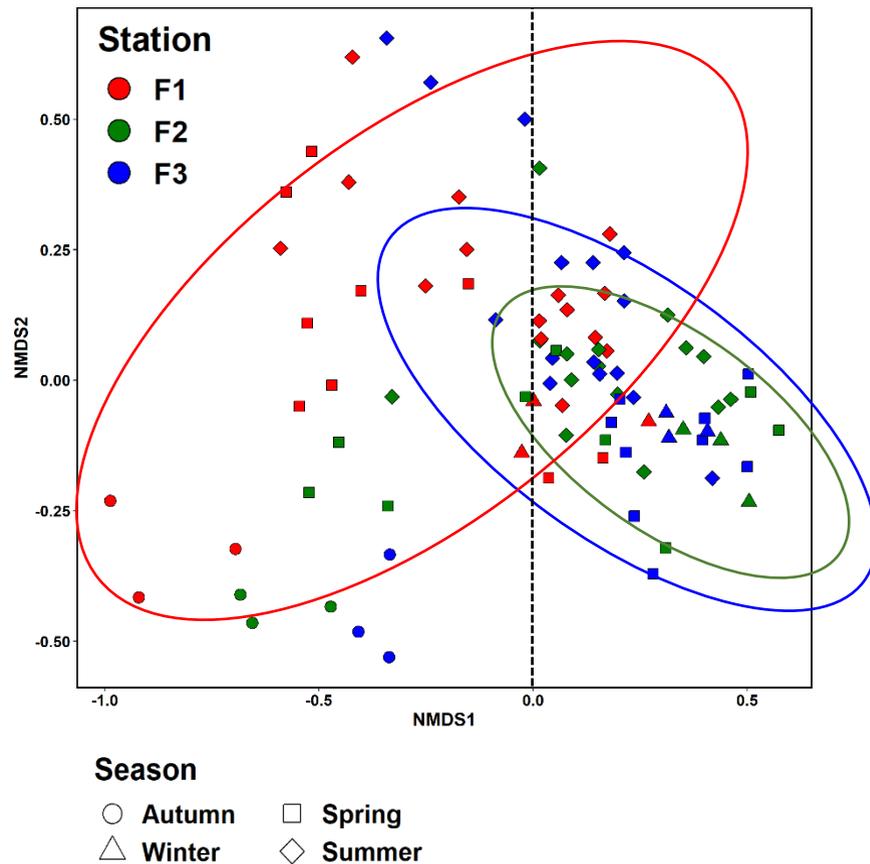
- (Other) Chlorophyta
- Chrysophyceae
- (Other) Haptophyta
- Pelagophyceae
- Prochlorococcus
- Synechococcus

- Cryptophyceae
- Prymnesiophyceae
- Dinoflagellates-2
- Dinoflagellates-1
- Prasinophytae
- Diatoms-2
- Diatoms-1

Group	r-coef	p-value	MAE	MSE	RMSE
<i>Synechococcus</i>	0.80	4.41E-07	0.18	0.05	0.22
<i>Cryptophyceae</i>	0.83	3.88E-07	0.12	0.02	0.14
Diatoms	0.74	7.73E-08	0.37	0.23	0.47
Dinoflagellates	0.56	2.03E-04	0.40	0.28	0.53

N = 25 - 39

Assemblage structure: variability and environmental predictors



Performance of RS-derived algorithms



Phytoplankton group (in situ pigments)	Satellite product	Performance metrics*		
		Bias	MAE	Overall Wins (%)
Pico	OCEANCOLOUR_ATL_BGC_L3_MY_009_113	1.29	1.67	27.27
	→ OCEANCOLOUR_GLO_BGC_L3_MY_009_103	1.20	1.50	54.55
	OCEANCOLOUR_GLO_BGC_L3_MY_009_107	1.11	1.73	18.18
	Abundance-based model with SST	1.24	1.68	0
Nano	OCEANCOLOUR_ATL_BGC_L3_MY_009_113	0.66	1.84	18.18
	OCEANCOLOUR_GLO_BGC_L3_MY_009_103	0.61	1.87	18.18
	OCEANCOLOUR_GLO_BGC_L3_MY_009_107	1.49	1.80	27.27
	→ Abundance-based model with SST	1.39	1.75	36.36
Micro	→ OCEANCOLOUR_ATL_BGC_L3_MY_009_113	0.60	1.67	36.36
	OCEANCOLOUR_GLO_BGC_L3_MY_009_103	0.57	1.88	9.09
	OCEANCOLOUR_GLO_BGC_L3_MY_009_107	0.71	1.74	18.18
	→ Abundance-based model with SST	0.70	1.77	36.36
Prokaryotes	OCEANCOLOUR_ATL_BGC_L3_MY_009_113	1.80	1.90	33.33
	→ OCEANCOLOUR_GLO_BGC_L3_MY_009_103	1.68	1.79	66.67
	Abundance-based model with SST	-	-	-
Diatoms	OCEANCOLOUR_ATL_BGC_L3_MY_009_113	0.35	2.88	9.09
	OCEANCOLOUR_GLO_BGC_L3_MY_009_103	0.34	2.98	18.18
	→ Abundance-based model with SST	0.63	1.91	72.73
Dinoflagellates	OCEANCOLOUR_ATL_BGC_L3_MY_009_113	5.99	5.99	8.33
	OCEANCOLOUR_GLO_BGC_L3_MY_009_103	5.16	5.16	5.88
	→ Abundance-based_model_with_SST	1.91	2.75	93.75

Number of valid matchups = 12 – 27. The best statistical results are highlighted in **bold**.

*Selected according to Seegers *et al.* (2018).

Conclusions

- Diatoms and dinoflagellates dominated at coastal stations, while cyanobacteria surpassed dinoflagellates further offshore
- Winter-spring: diatom dominance; summer-autumn: cyanobacteria and dinoflagellate dominance
- UI, NO₂ and SPM were the best predictors of variability in the structure of the phytoplankton assemblage
- RS-based algorithms need further refinement and validation to be applied at regional scales

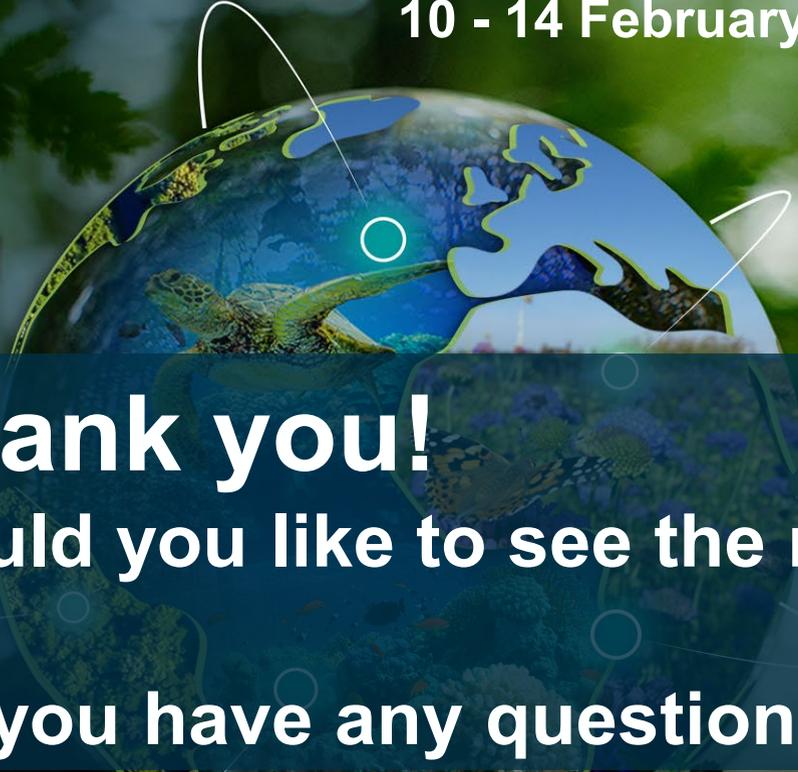
Next steps

- Estimate NO_3^- concentrations in the study area from *in situ*/satellite-retrieved water temperature, SiO_4^{4-} , and/or Chl-a concentrations using empirical regressions
- Apply multivariate analysis with model-derived NO_3^- concentrations
- Explore specific bloom events

Recommendations for conference organizers:

- Provide travel funding/scholarships for participants from developing countries
- Implement real-time feedback mechanisms (e.g., post-session surveys)
- Promote Q&A sessions/round tables after each session

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Thank you!
Would you like to see the reference list?
Do you have any questions?

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