

Using IFCB to monitor filamentous cyanobacteria in the Baltic Sea

Automated data pipeline and CNN based classification system

Kaisa Kraft, Otso Velhonoja, Jukka Seppälä, Heidi Hällfors, Sanna, Pasi Ylöstalo, Sami Kielosto, Sirpa Lehtinen, Johanna Oja, Timo Tamminen
Finnish Environment Institute SYKE, Marine Research Centre

25.8.2022



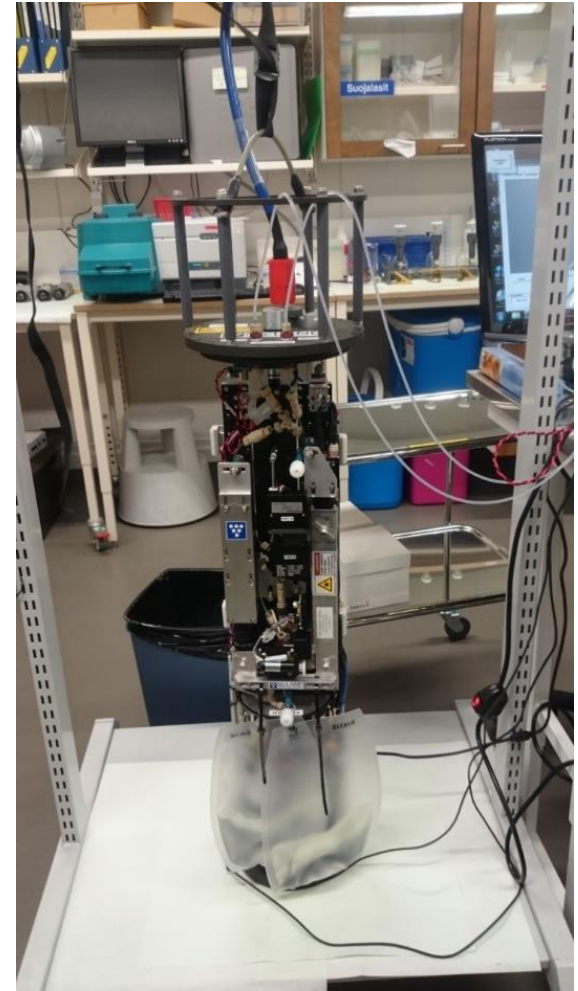
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FASTVISION



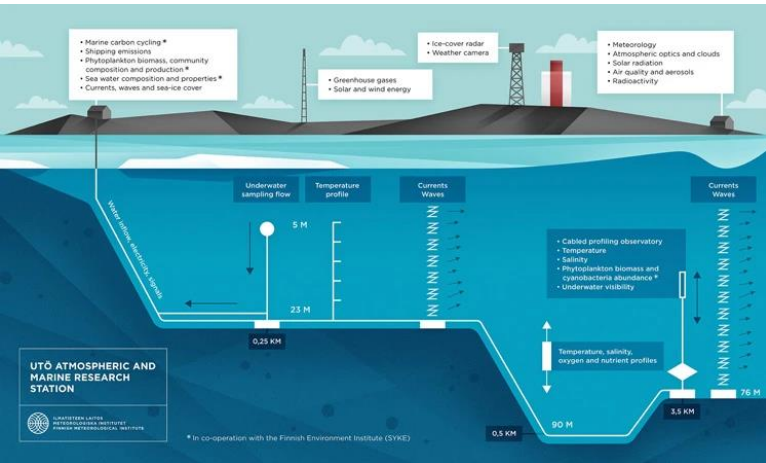
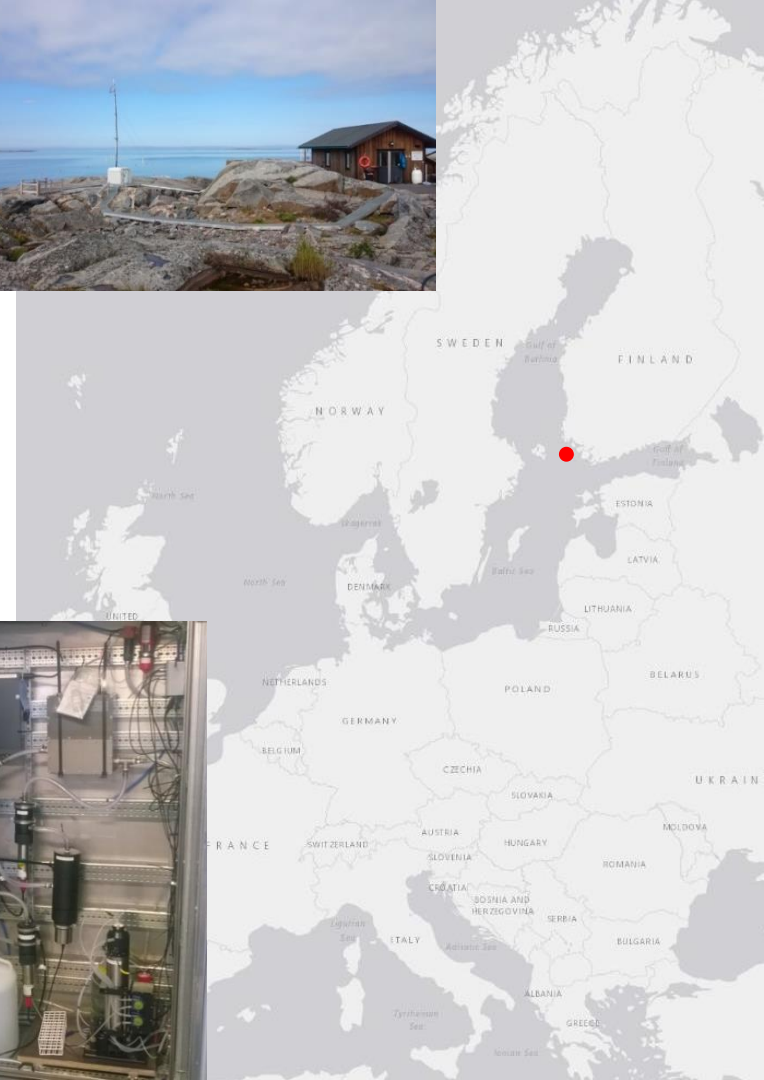
Imaging FlowCytobot - IFCB

- Imaging flow cytometer
- Takes images of phytoplankton cells and colonies inside the size range of 10-150 μm
- Can be operated remotely mounted on a flow through system at a research station (like in Utö), ship of opportunity, put to measure directly to the sea or used in a laboratory environment etc..
- Takes a sample of 5ml with approx. 20 min interval
- The camera is triggered by chlorophyll-*a* or scatter
- Even as many as ~30 000 high resolution images / hour
- 150 μm mesh in IFCB inlet to prevent it from clogging
- Data analyzed with image recognition algorithm

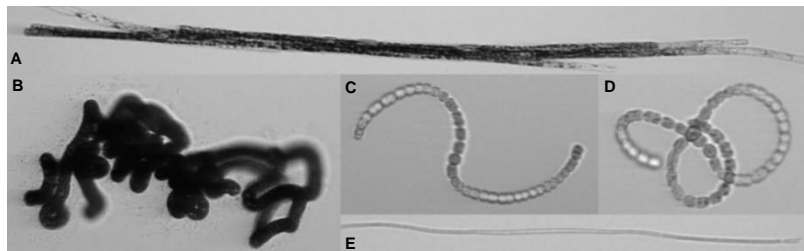


UTÖ Marine Research Station joint with Finnish Meteorological Institute

- Underwater pump with inlet at 5 m depth, 250 m offshore
- Water distributed to different sensors inside the station cabin
- Represents pelagial community of a mixed surface layer in brackish environment (salinity ~ 6 psu)
- Multiple parallel measurements from sea to atmosphere
- Continuous imaging flow cytometer observations from multiple years
- Light microscopy samples of cyanobacteria bloom in 2018



Cyanobacteria blooms in the Baltic Sea



Common filamentous species

- *Aphanizomenon flosaquae* (A)
- *Dolichospermum* spp. (C,D)
- *Nodularia spumigena* (B)
- Oscillatoriales (E)

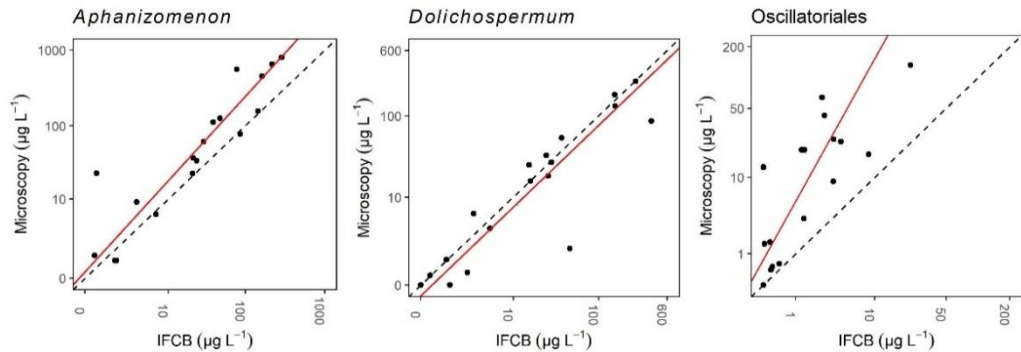
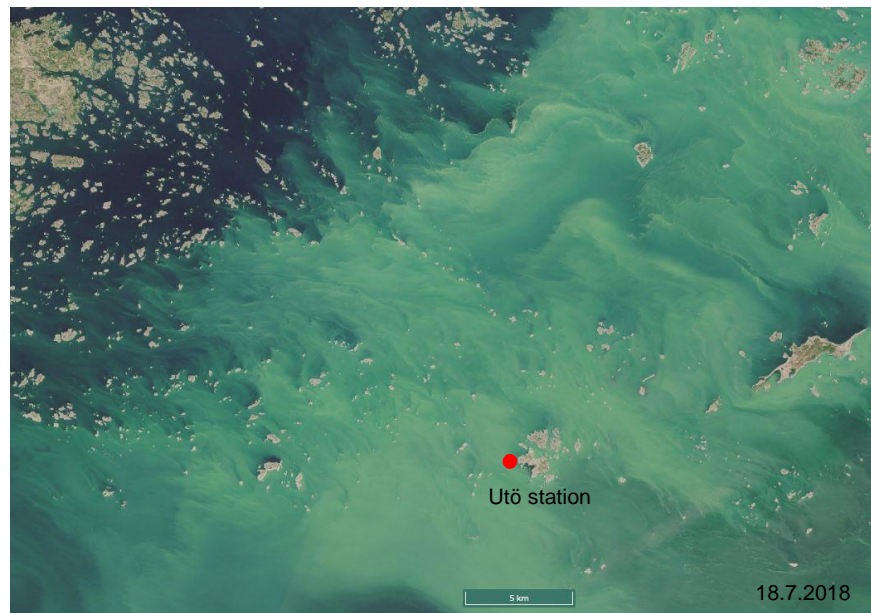


Image analysis was performed using the publicly available code by H. Sosik et al. (Sosik & Olson 2007)

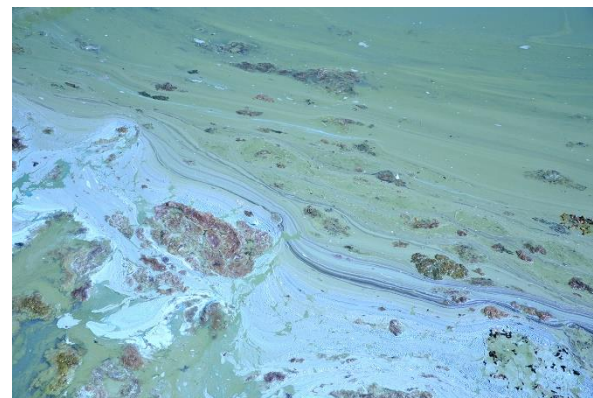
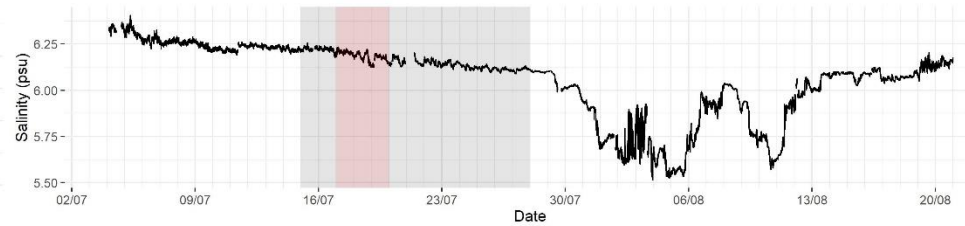
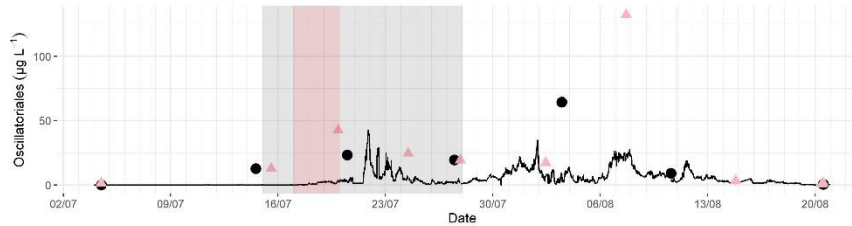
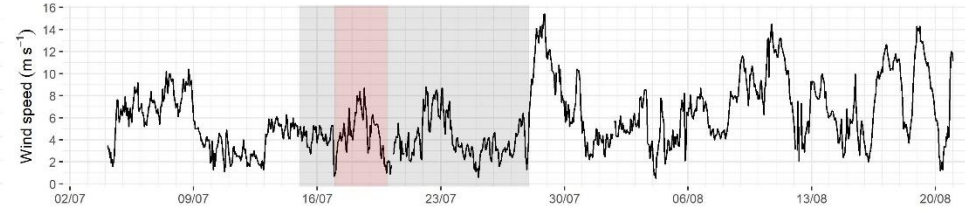
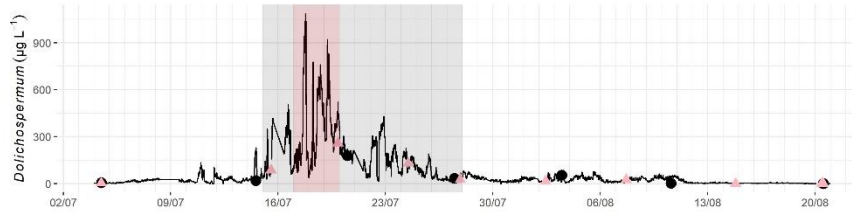
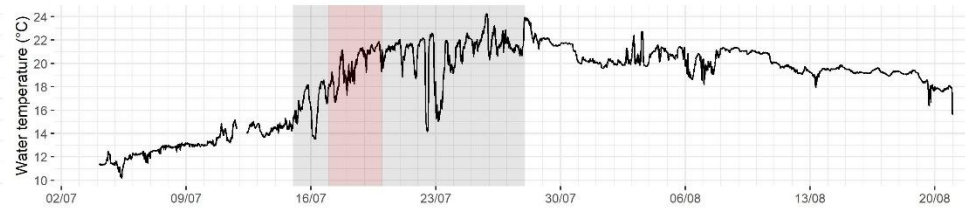
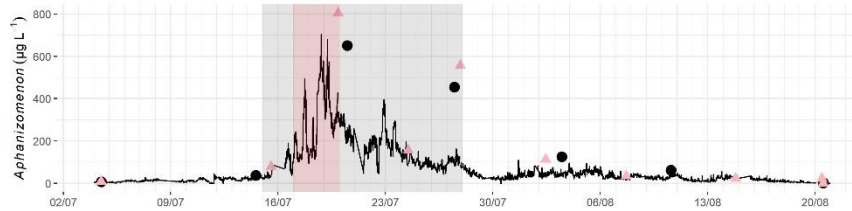


Photo by Lauri Laakso

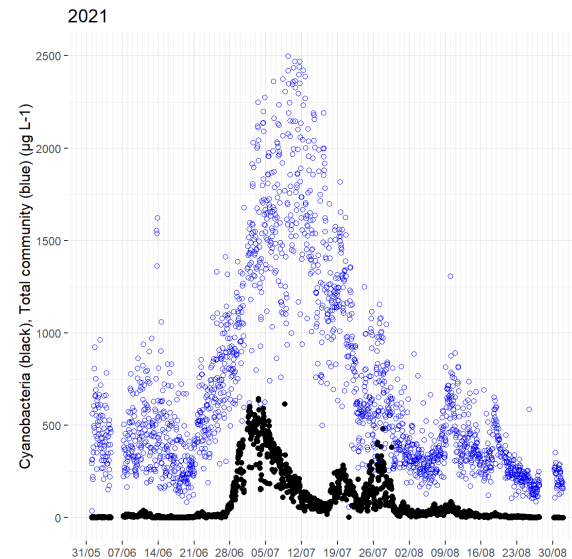
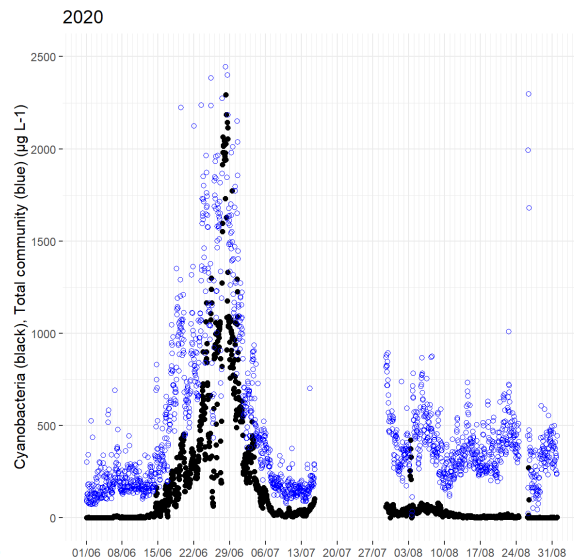
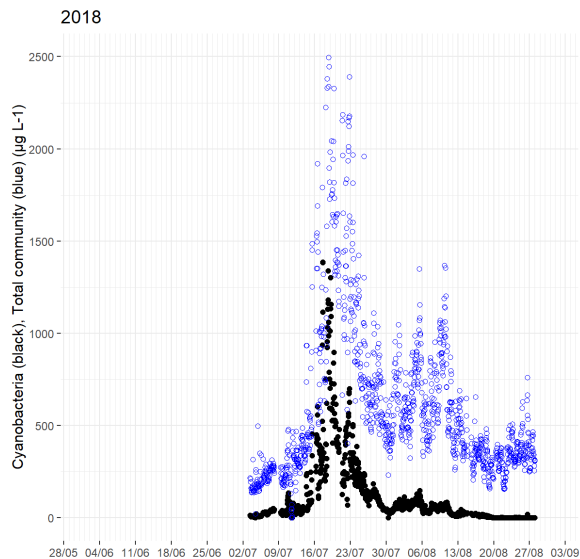
High frequency phytoplankton observations connected with changes in the environment can help to understand mechanisms of the blooms



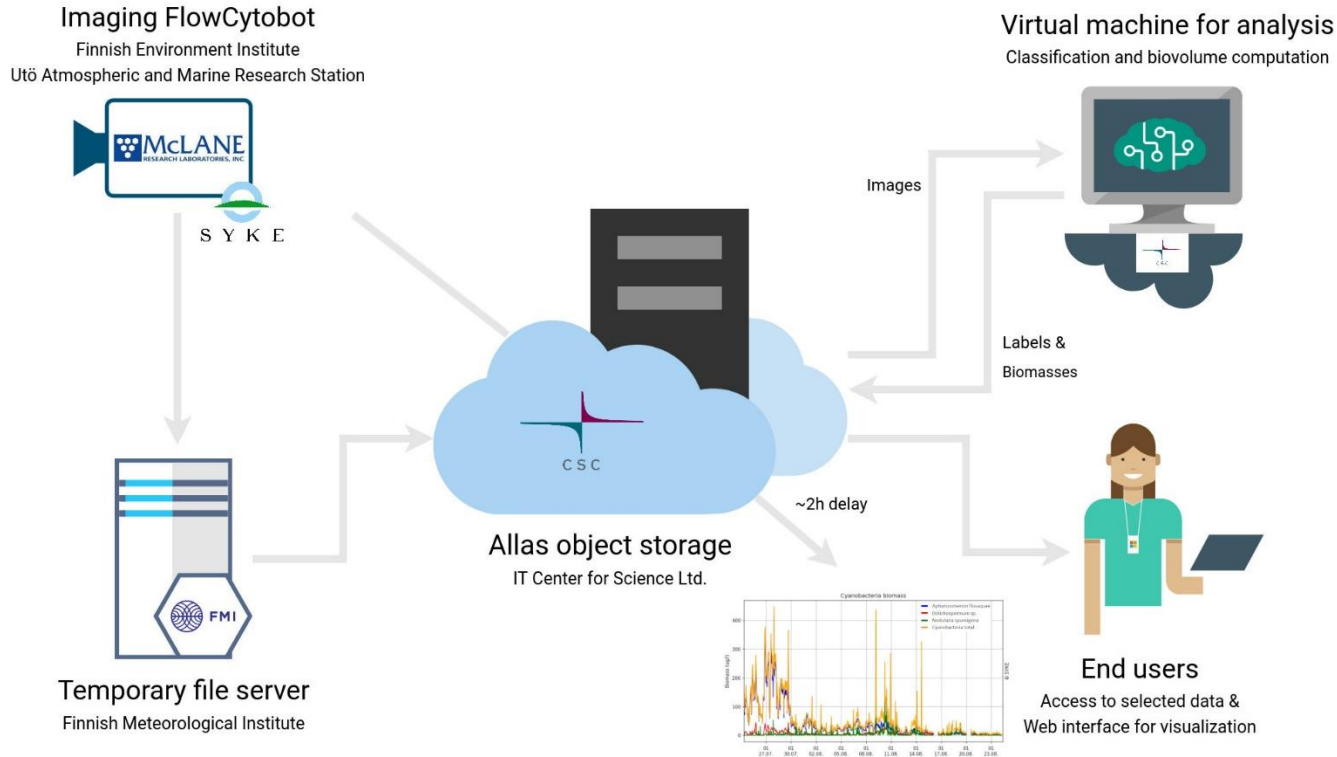
bloom period highlighted in grey (biomass > approximately $100 \mu\text{g L}^{-1}$)
bloom peak period highlighted in red (the days with the highest peaks)
light microscopy counts ● flow-through ▲ pump inlet

-> species dynamics were connected to changes in the water mass

Cyanobacteria blooms in 2018, 2020, 2021



Near-real time classification via Convolutional Neural Networks



- Training data set
- <http://doi.org/10.23728/b2share.abf913e5a6ad47e6baa273ae0ed6617a>
- Evaluation data set
- <http://doi.org/10.23728/b2share.7c273b6f409c47e98a868d6517be3ae3>
- Visualization still under development
- <https://plankton.live/>
- <https://ifcb.plankton.live/timeline?dataset=uto>
- <https://swell.fmi.fi/hab-info/>

Classifier performance

- Kraft et al. 2022 Front Mar Sci
- Weighted F1-score of test data of our labeled image data set 0.95
- Weighted F1-score of our evaluation data (59 natural samples annotated entirely) 0.83
- Class-specific thresholds are used for filtering out the unidentifiable images, determined with our test set complemented with unidentifiable images (initial situation -> to be finetuned as data accumulates)
- ND in the Evaluation Data means “Not Determined”, the metrics were not calculated for classes with <10 images

Class / taxonomic group	Training	Validation	Test				2021 data			
	N	Thre	N	Pr	Re	F1	N	Pr	Re	F1
Oscillatoriales	2664	0.31	888	0,99	1,00	0,99	3893	0,98	0,98	0,98
<i>Monoraphidium contortum</i>	196	0.69	66	0,98	0,98	0,98	439	0,99	0,96	0,97
<i>Skeletonema marinoi</i>	2477	0.46	825	1,00	0,99	0,99	7402	0,99	0,94	0,97
<i>Heterocapsa triquetra</i>	1966	0.39	655	0,98	0,97	0,97	2267	0,92	0,95	0,94
<i>Cryptophyceae / Teleaulax sp.</i>	4098	0.53	1366	0,96	0,97	0,96	16952	0,97	0,90	0,93
<i>Aphanizomenon flosaquae</i>	4193	0.24	1398	0,97	1,00	0,98	1849	0,87	0,98	0,92
<i>Peridiniella catenata</i> chain	116	0.7	38	0,97	1,00	0,99	89	0,99	0,87	0,92
<i>Dolichospermum sp. / Anabaenopsis sp.</i>	7368	0.38	2456	0,98	0,99	0,98	790	0,88	0,96	0,92
<i>Pauliella taeniata</i>	71	0.62	24	1,00	0,96	0,98	56	0,96	0,86	0,91
<i>Oocystis sp.</i>	505	0.5	169	0,88	0,93	0,90	161	0,91	0,89	0,90
<i>Mesodinium rubrum</i>	679	0.44	227	0,96	0,95	0,96	560	0,92	0,86	0,89
<i>Melosira arctica</i>	26	0.3	8	0,73	1,00	0,84	58	0,85	0,91	0,88
<i>Dolichospermum sp. / Anabaenopsis sp. coiled</i>	1502	0.41	501	0,93	0,96	0,95	70	0,74	0,99	0,85
<i>Eutreptiella sp.</i>	1348	0.43	450	0,95	0,94	0,94	1678	0,90	0,76	0,83
<i>Licmophora sp.</i>	44	0.43	15	1,00	0,80	0,89	78	0,90	0,77	0,83
<i>Nodularia spumigena</i>	101	0.32	34	0,80	0,94	0,86	62	0,80	0,85	0,83
<i>Heterocapsa rotundata</i>	368	0.56	123	0,84	0,90	0,87	2609	0,89	0,70	0,78
<i>Ceratoneis closterium</i>	27	0.41	9	1,00	1,00	1,00	75	0,68	0,91	0,78
<i>Peridiniella catenata</i> single	539	0.52	180	0,89	0,97	0,93	222	0,75	0,81	0,78
Pennales thick	126	0.37	42	0,93	0,88	0,90	1088	0,72	0,85	0,78
<i>Thalassiosira levanderi</i>	1522	0.63	508	0,95	0,95	0,95	2008	0,87	0,68	0,77
<i>Chaetoceros sp. chain</i>	829	0.51	277	0,93	0,95	0,94	693	0,76	0,77	0,76
Centrales	288	0.51	96	0,98	0,89	0,93	92	0,77	0,68	0,72
<i>Dinophysis acuminata</i>	130	0.68	44	0,98	0,91	0,94	17	0,79	0,65	0,71
Pennales thin	469	0.29	156	0,96	0,99	0,97	334	0,61	0,84	0,71
<i>Cyclotella choctawhatcheeana</i>	61	0.47	21	0,89	0,81	0,85	199	0,92	0,57	0,71
Gymnodiniales	41	0.29	14	0,92	0,86	0,89	38	0,78	0,64	0,70

Table contains results of evaluation data for F1 > 0,7

Group specific confusion of our evaluation data

Actual label

Cyanophyceae	0.94	0	0	0	0	0	0	0.06	
Cryptophyceae	0	0.89	0	0	0	0	0	0.11	
Euglenophyceae	0	0.02	0.76	0.01	0	0	0	0.22	
Dinophyceae	0	0	0.01	0.75	0	0	0	0.24	
Bacillariophyceae	0.01	0	0	0	0.86	0	0	0.13	
Chrysophyceae	0	0	0	0	0	0.51	0	0.49	
Chlorophyta	0	0	0	0	0	0	0.34	0.65	
Ciliophora	0	0	0	0.02	0.03	0	0	0.76	0.19

Predicted label

What can we identify?

~50 categories, ~60000+ annotated images

Taxons / groups so far identified with IFCB from the Baltic Sea (work ongoing)

Cyanophyceae

Aphanothece paralleliformis

Chroococcales

Chroococcus spp.

Merismopedia spp.

Snowella spp. / *Woronichinia* spp.

Oscillatoriales

Aphanizomenon flosaquae

Aphanizomenon spp.

Dolichospermum spp. / *Anabaenopsis* spp.

Nodularia spumigena

Cryptophyceae

Cryptomonadales

Teleaulax spp.

Dinophyceae

Prorocentrum cordatum

Dinophyceae (under & over 20µm)

Dinophysis acuminata

Dinophysis norvegica

Gymnodiniales

Gymnodinium spp.

Gymnodinium like cells

Gyrodinium spp.

Heterocapsa rotundata

Heterocapsa triquetra

Amylax triacantha

Gonyaulax verior

Peridiniella catenata

Protoceratium reticulatum

Chrysophyceae

Dinobryon balticum

Uroglenopsis spp.

Apedinella radians

Pseudopedinella spp.

Diatomophyceae

Chaetoceros spp.

Chaetoceros danicus

Chaetoceros similis

Chaetoceros subtilis

Chaetoceros thronsenii

Coscinodiscus granii

Cylindrotheca closterium

Cyclotella choctawhatcheana

Melosira arctica

Skeletonema marinoi

Thalassiosira baltica

Thalassiosira levanderi

Diatoma tenue

Licmophora spp.

Nitzschia paleacea

Pauliella taeniata

Euglenophyceae

Eutreptiella spp.

Chlorophyta

Cymbomonas tetramitiformis

Pyramimonas spp.

Chlorococcales

Scenedesmus spp. / *Acutodesmus* spp.

/ *Desmodesmus* spp.

Monoraphidium contortum

Binuclearia lauterbornii

Oocystis spp.

Katablepharis remigera

Flagellates

Nanoplankton

Ciliates

Strombidium spp.

Mesodinium rubrum

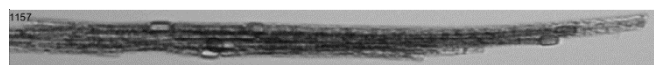
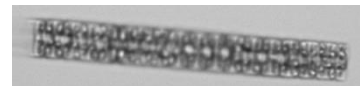
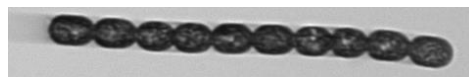
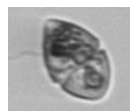
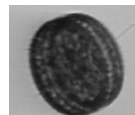
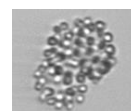
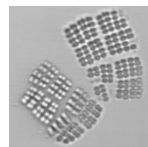
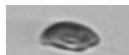
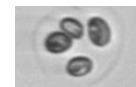
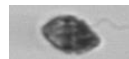
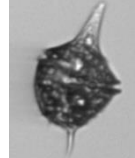
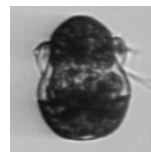
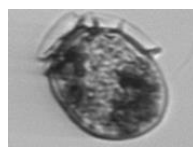
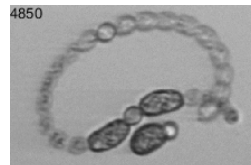
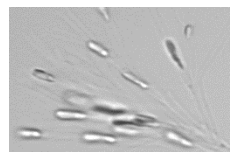
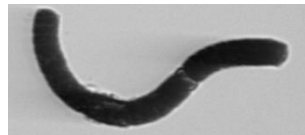
Additional categories include eg.

akinetes

heterocytes

resting stages

cysts



References and sources

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