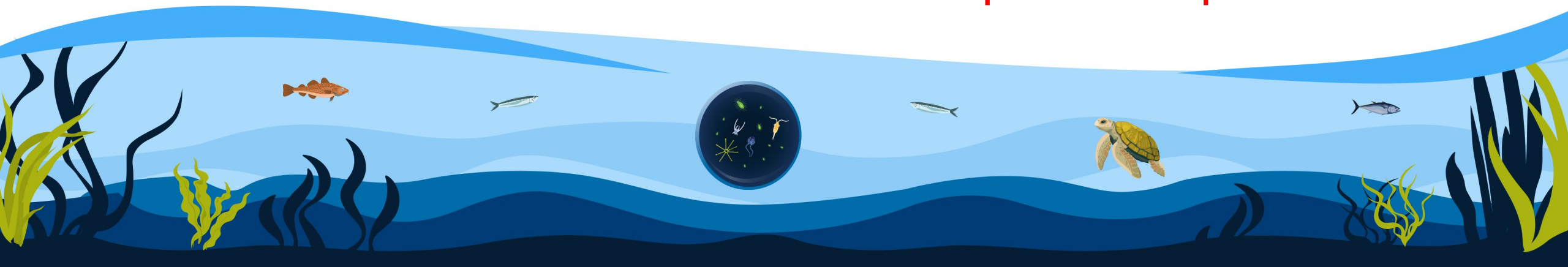


# Revising Carbon Uptake Estimates in the European Arctic with a regional satellite algorithm and BGC-Argo data

Aleksandra Cherkasheva, Artur Palacz, Rustam Manurov, Piotr Kowalczyk,

Alexandra Loginova, Astrid Bracher

25 November 2<sup>nd</sup> Ocean Carbon from Space workshop



# Marine Organic Carbon Atlas (MOCA) Pilot Demonstration for the Arctic



C uptake



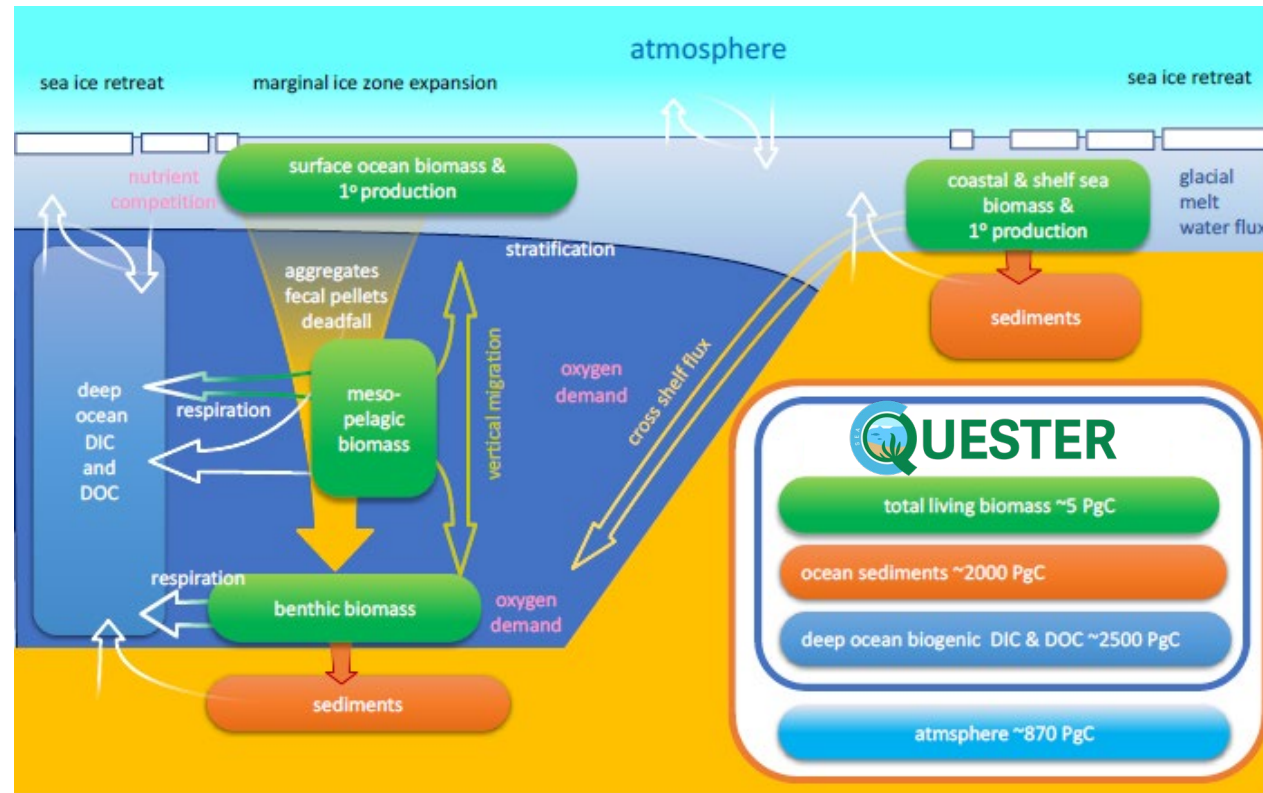
C export



C recycling



C burial



From: [www.sea-quester.eu](http://www.sea-quester.eu)



[http://www.ioccp.org/images/D2backgroundDoc/IOCR\\_WG\\_Report\\_2021.pdf](http://www.ioccp.org/images/D2backgroundDoc/IOCR_WG_Report_2021.pdf)

Contact: Artur Palacz  
palacz@iopan.pl



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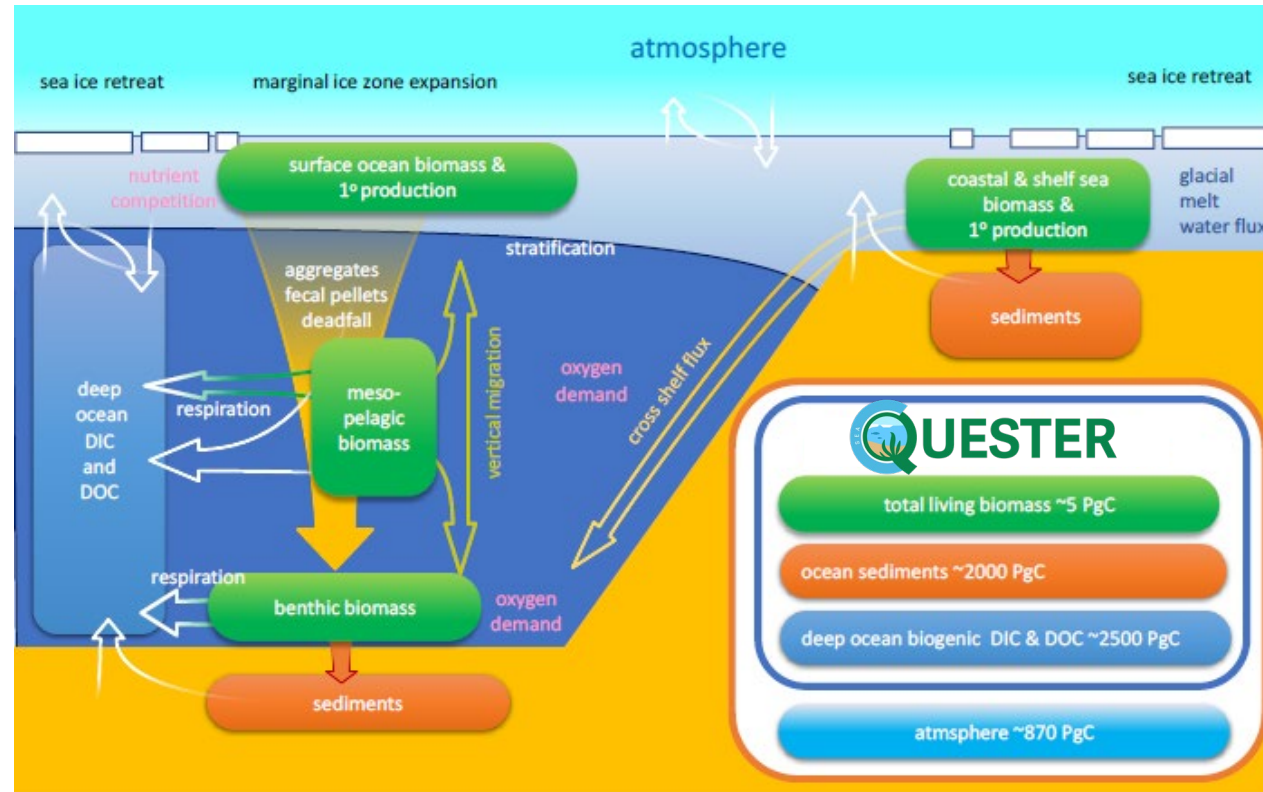
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# Greenland Sea Primary Production (PP) algorithm development

$$P = (12/4.6)CHL_{tot} \overline{PAR(0^+)} \overline{a^*} \overline{\varphi_{\mu}}$$

*Morel(1991)*



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- Chlorophyll a
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- PI parameters (from Bouman et al. 2018)
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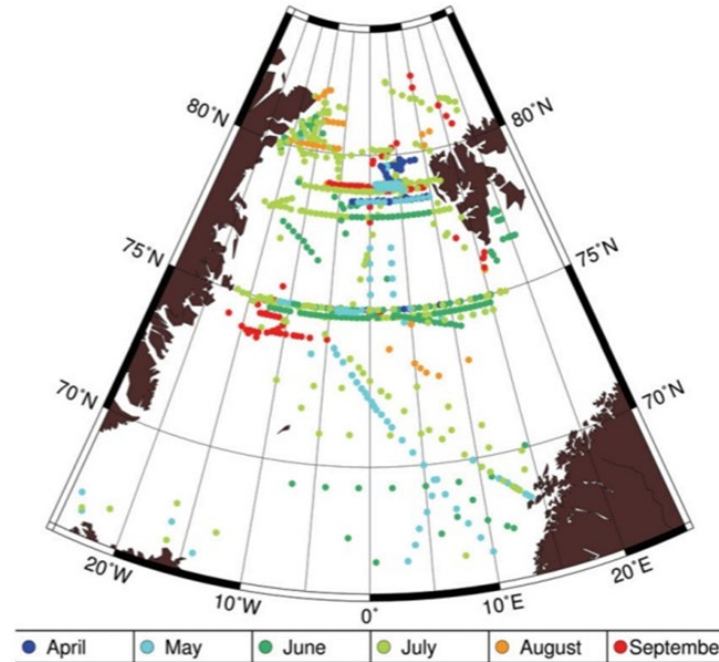
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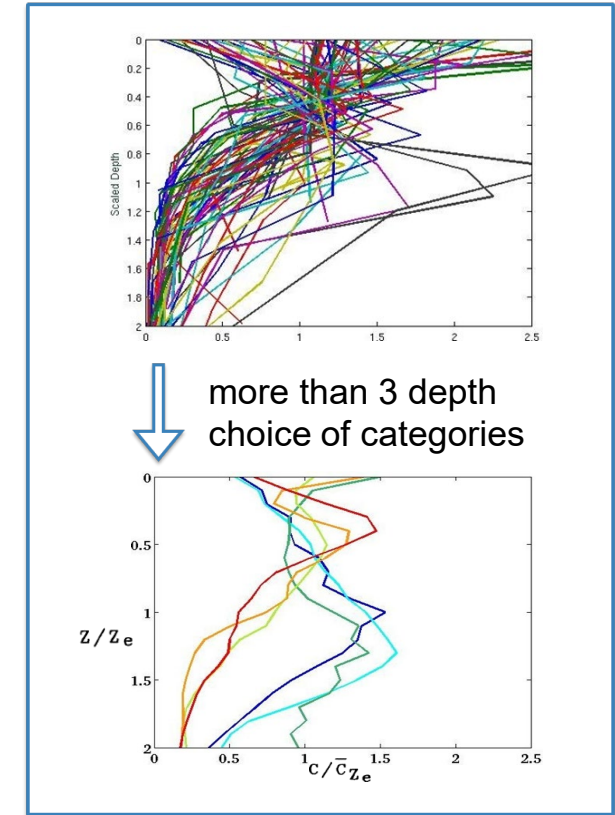
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Example: CHL profile parameterization

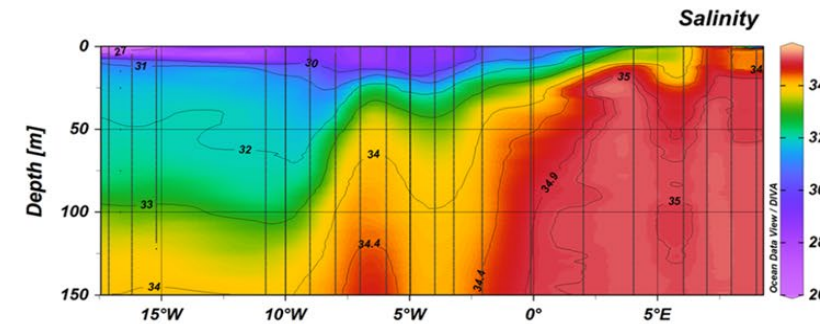
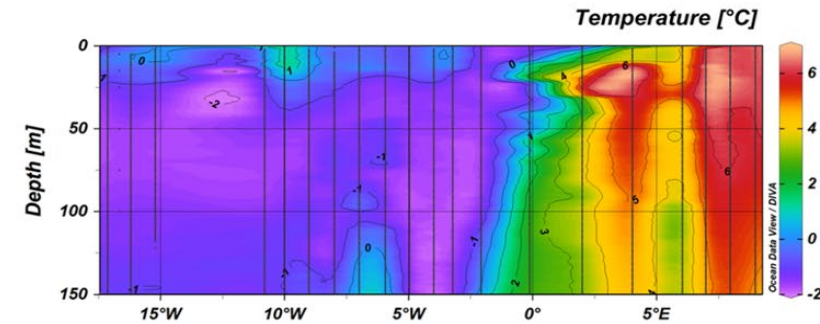
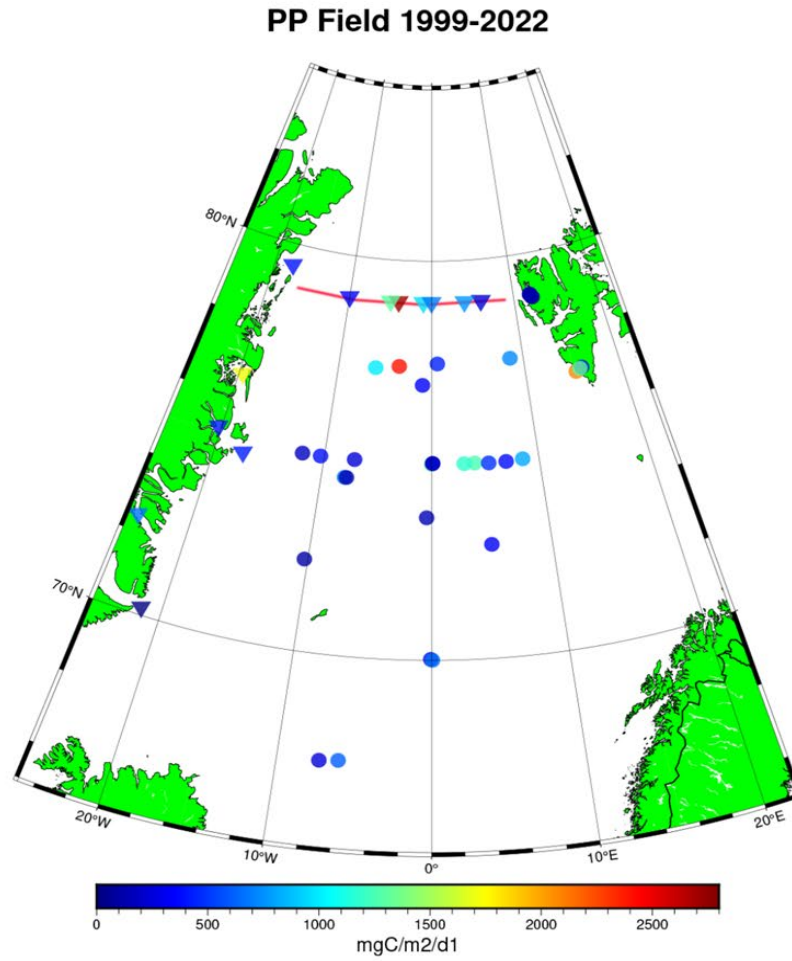


- 1680 data profiles
- years from 1957



Cherkasheva et al. (2013)

# Validation: PP field data





# Best performing model setup

Type of CHL	Type of PAR	CHL vertical profile	PAR/PUR	Integration depth
[3] - GlobColour Level 3 CHL	[1] - NOAA/NCEP Reanalysis PAR	[0] - Global CHL profile following Morel and Berthon (1988)	[0] - no coefficient applied to PAR	[0] - profiles integrated till euphotic layer depth
[4] - Copernicus-GlobColour Level 4 CHL;	[2] - EUMETSAT Level 2 OLCI PAR	[1] - Local CHL profile following <a href="#">Cherkasheva et al. (2013)</a>	[1] - PAR converted to local PUR	[1] - profiles integrated till productive layer depth
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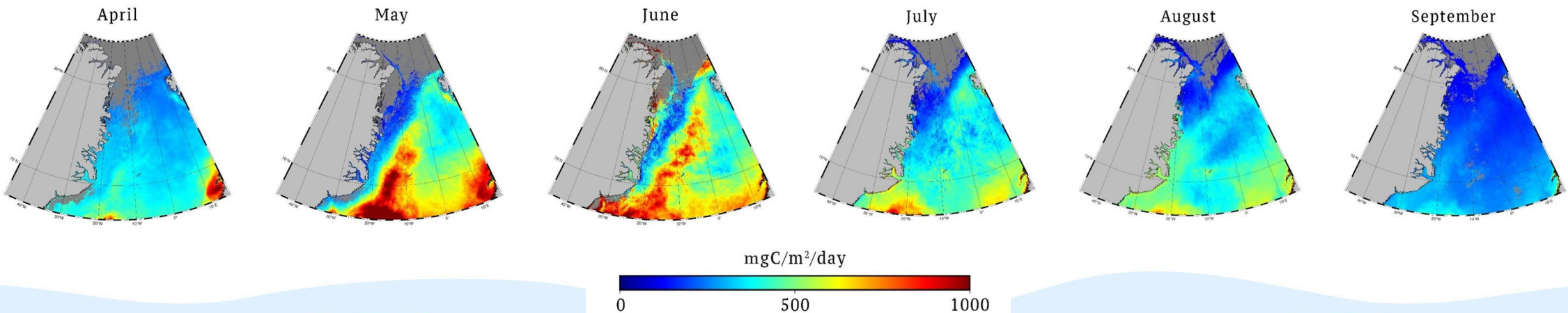
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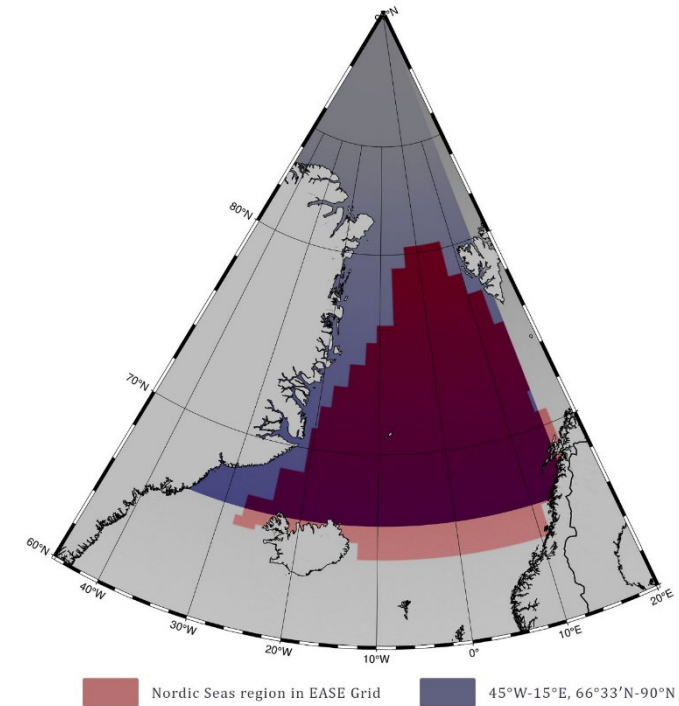
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Monthly averaged PP, 1998-2022



# Results: Greenland Sea Basin estimates

	This study	Hill et al. (2013)	Arrigo and van Dijken (2015)	Ardyna et al (2013)
period	1998-2022	1998-2007	1998-2012	1998, 2007
annual (TgC/year)	330-346	308	136	228-230
monthly range(TgC/month)	33-77	13-122	ND	ND



## Why is it larger? possible reasons

- account for local and not Arctic dataset (5 points vs 45 points),
- more spatial coverage in the northern part,
- account for CHL maximum



# Results: Greenland Sea Basin estimates

	Source	Region	Period	Annual (TgC/year)	Month (TgC/month)					
					Apr	May	Jun	Jul	Aug	Sep
1	<a href="#">Hill et al. (2013)</a> with SCM	Nordic Seas region in EASE Grid	1998-2007	308	33.4	122.5	42.3	27.5	13.4	44.5
2	<a href="#">Arrigo and van Dijken (2015)</a>	45°W-15°E, 66°33'N-90°N	1998-2012	136.3	ND	ND	ND	ND	ND	ND
3	<a href="#">Ardyna et al. (2013)</a>	Greenland-Norwegian Seas	1998	227.9	ND	ND	ND	ND	ND	ND
			2007	230.8	ND	ND	ND	ND	ND	ND
4	This study (related to #1)	Nordic Seas region in EASE Grid	1998-2007	344.1 (5.4)	45.3	77.9	73.7	60.3	54.1	32.7
5	This study (related to #2)	45°W-15°E, 66°33'N-90°N	1998-2012	340.0 (10.6)						
6	This study (related to #3)	45°W-15°E, 66°33'N-90°N	1998	333.1 (29.8)						
			2007	333.7 (1.5)						
7	This study (own estimates)	Nordic Seas region in EASE Grid	1998-2022	346.6 (2.4)	44.4	77.1	76.0	60.6	55.2	33.2
8	This study (own estimates)	45°W-15°E, 66°33'N-90°N	1998-2022	342.1 (9.6)	42.8	69.7	74.1	62.3	58.3	35.0

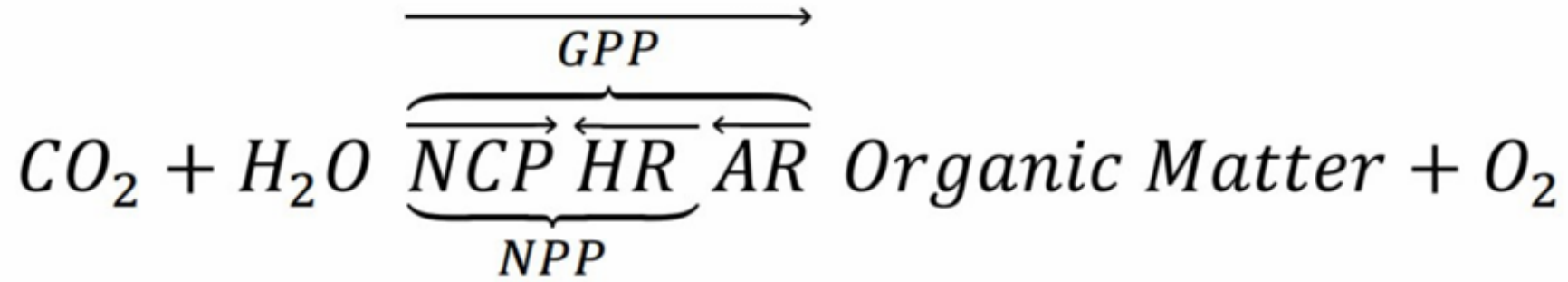
The average between the two setups of models [4,2,1,1,1] and [3,2,1,1,0] is given, the standard deviation is the number in the brackets. Monthly values from [Hill et al. \(2013\)](#) are in italics as we have calculated them ourselves from [Hill et al. \(2013\)](#) averages without SCM using a method given in the source. ND, no data found in the source.

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# Different types of production



**GPP - Gross Primary Production**

**NPP - Net Primary Production**

**NCP - Net Community Production**

**AR - Autotrophic Respiration**

**HR - Heterotrophic Respiration**

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**C uptake**



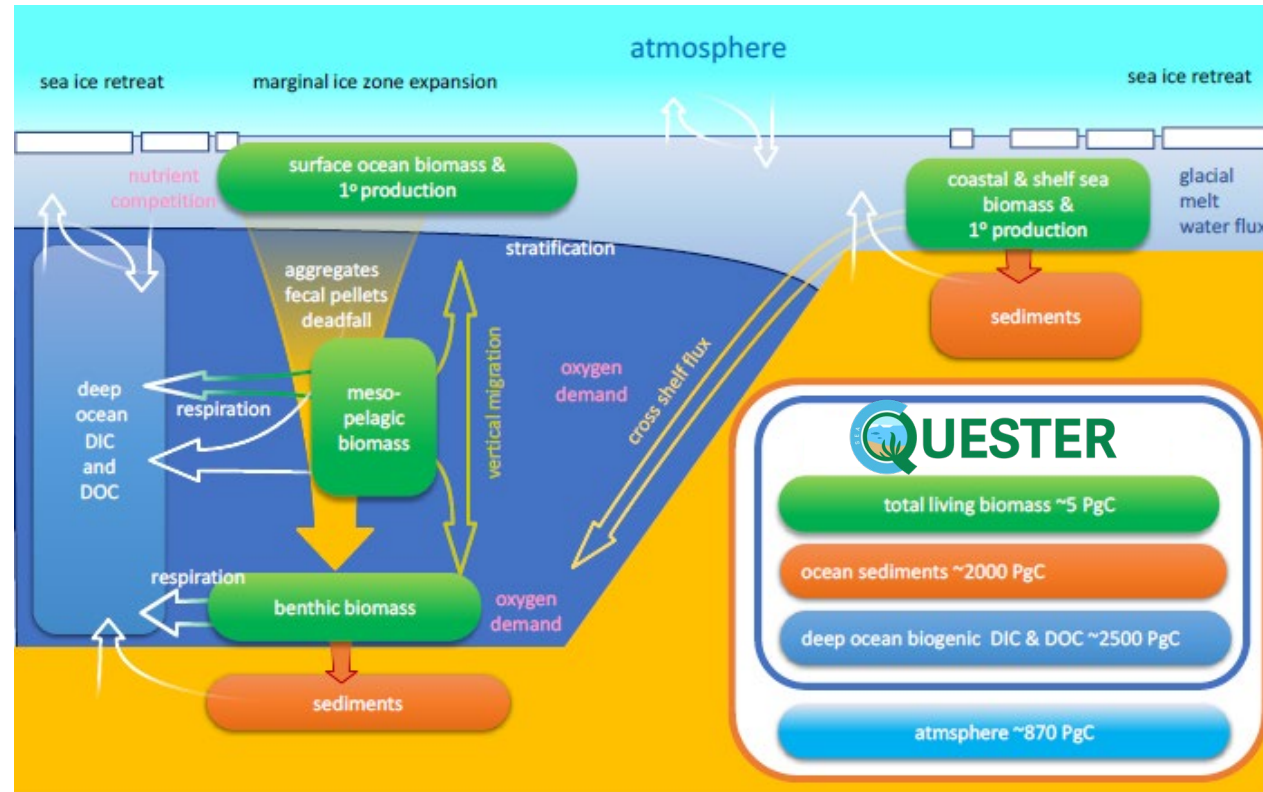
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From: [www.sea-quester.eu](http://www.sea-quester.eu)



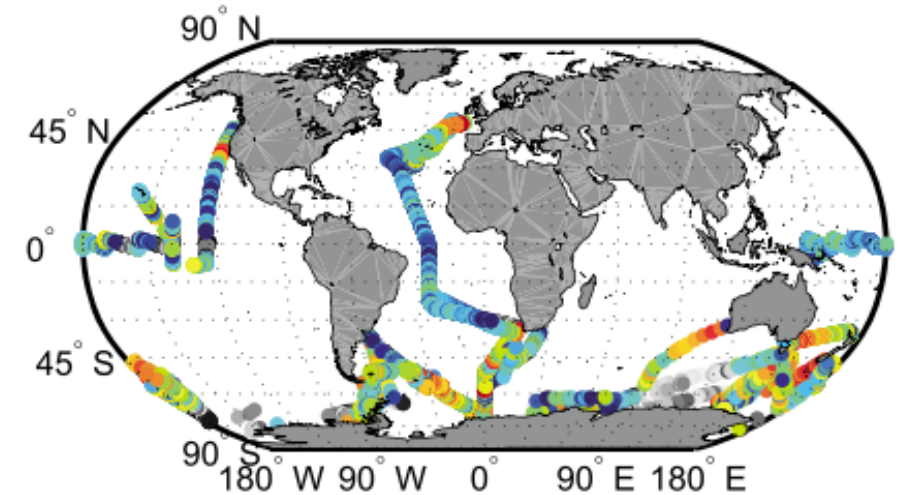
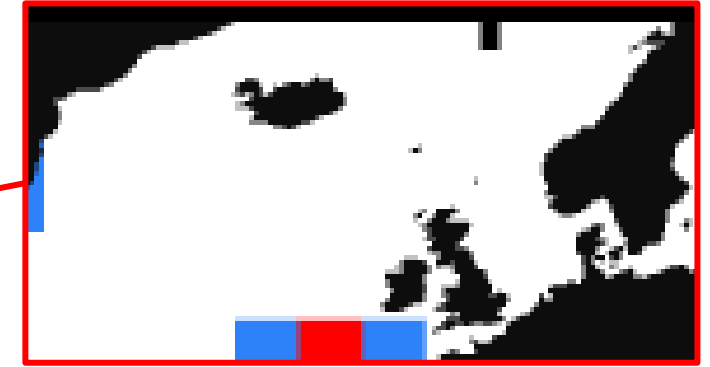
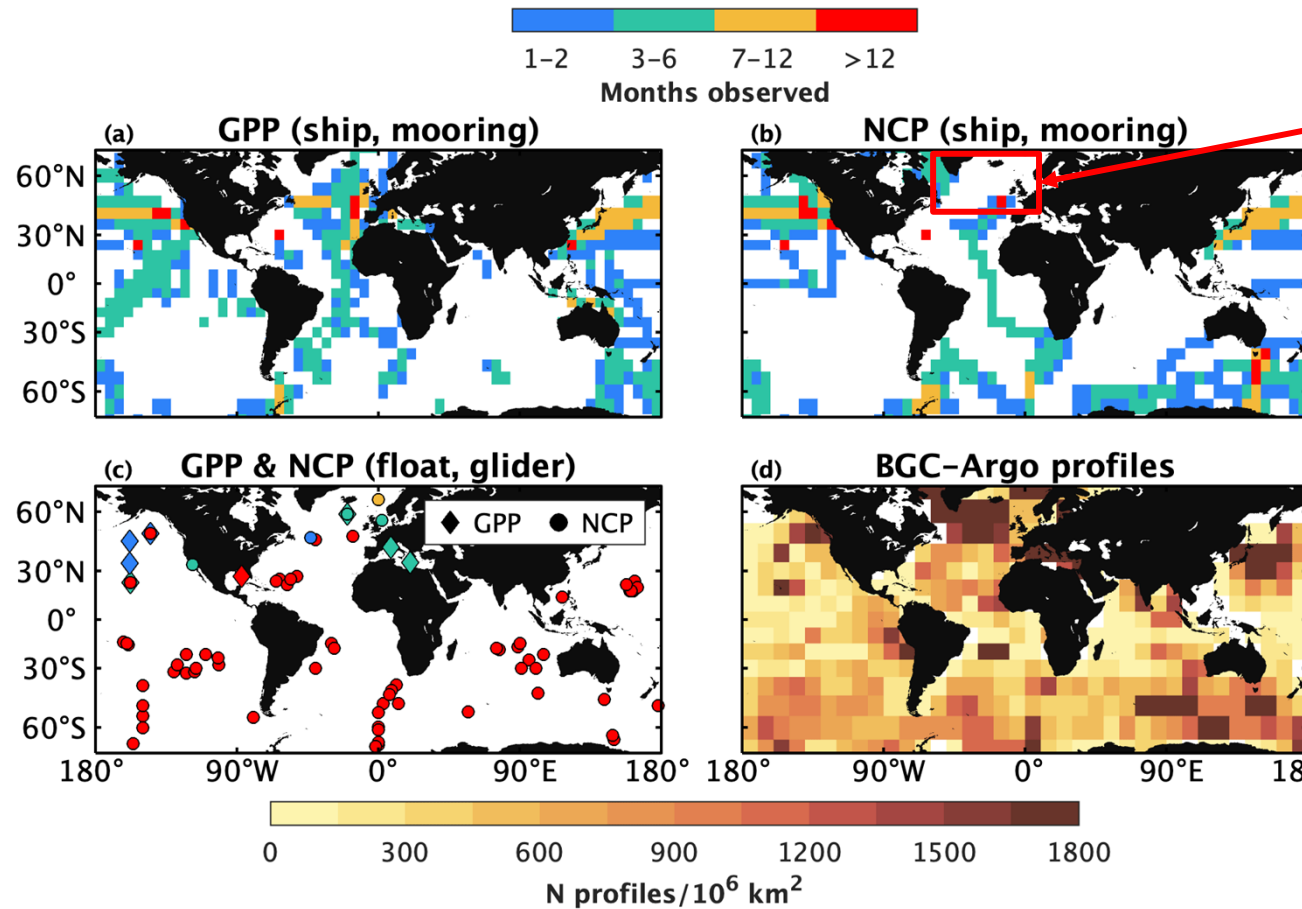
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# NCP estimates: main challenges



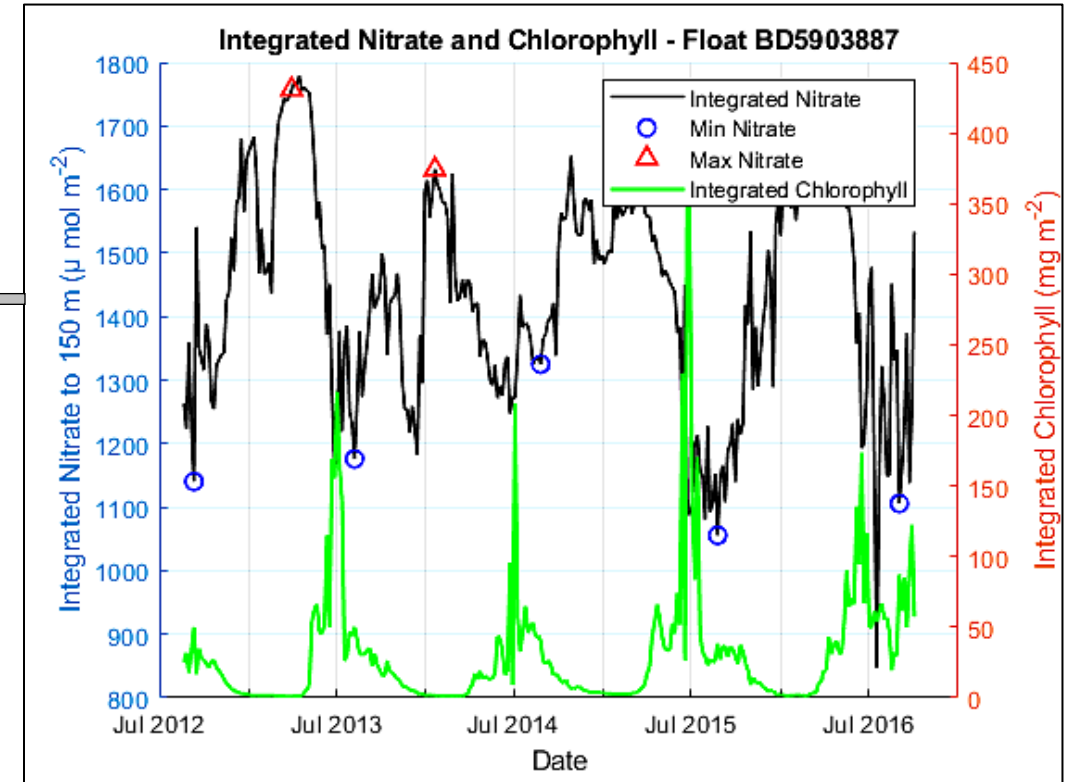
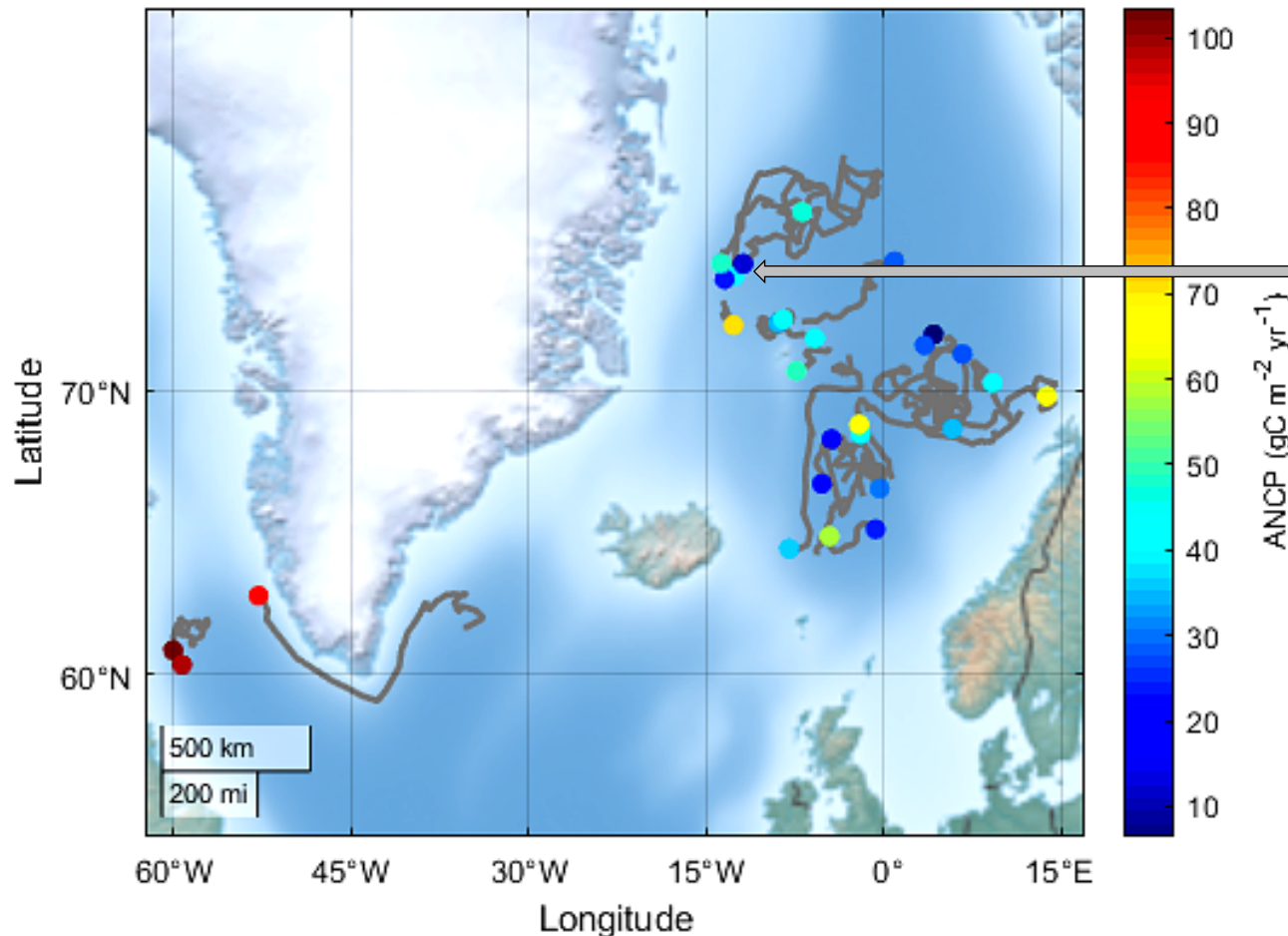
# NCP estimates: main challenges

Method	Field Data Needed	Temporal scale	Cons
In Situ Incubations	O <sub>2</sub> or DIC changes in light/dark bottles	Daily	high sampling effort, short-term estimates
Oxygen-to-Argon (O <sub>2</sub> /Ar) Method	O <sub>2</sub> /Ar ratio, temperature, wind speed	Daily to weekly	Requires specialized equipment, surface-limited
Oxygen Mass Balance (BGC-Argo)	O <sub>2</sub> profiles, wind speed, MLD	Weekly to annual	Sensitive to gas exchange errors
Nitrate sensors (BGC-Argo)	nitrate profiles, MLD	Weekly to annual	Ignores regenerated production, weekly to annual estimates (not daily)
DIC Mass Balance	pCO <sub>2</sub> (converted to DIC), salinity, temperature	Seasonal to annual	mostly surface-limited, sensitive to gas exchange errors
Nitrate Drawdown (from water samples)	Nitrate profiles	Seasonal to annual	Ignores regenerated production (leads to underestimation), could be challenging to estimate winter nitrate levels
Triple Oxygen Isotopes	Isotopic O <sub>2</sub> samples	Daily to weekly	Lab processing effort unknown to me, specialized equipment needed
Optical Sensors (BGC-Argo) (?)	Backscatter, fluorescence	Daily	Indirect, calibration needed
Thorium-234 Method (attached to sinking particles)	<sup>234</sup> Th and POC profiles	Weekly to monthly	Requires radionuclide handling, gives a component of NCP usually used as an additional method in NCP intercomparison
Sediment Traps	POC flux data	Seasonal to annual	Expensive, gives a component of NCP usually used as an additional method in NCP intercomparison
Various sensors on gliders (oxygen, pCO <sub>2</sub> )	same as p.4	Weekly to seasonal	same as p.4

[link to table](#)

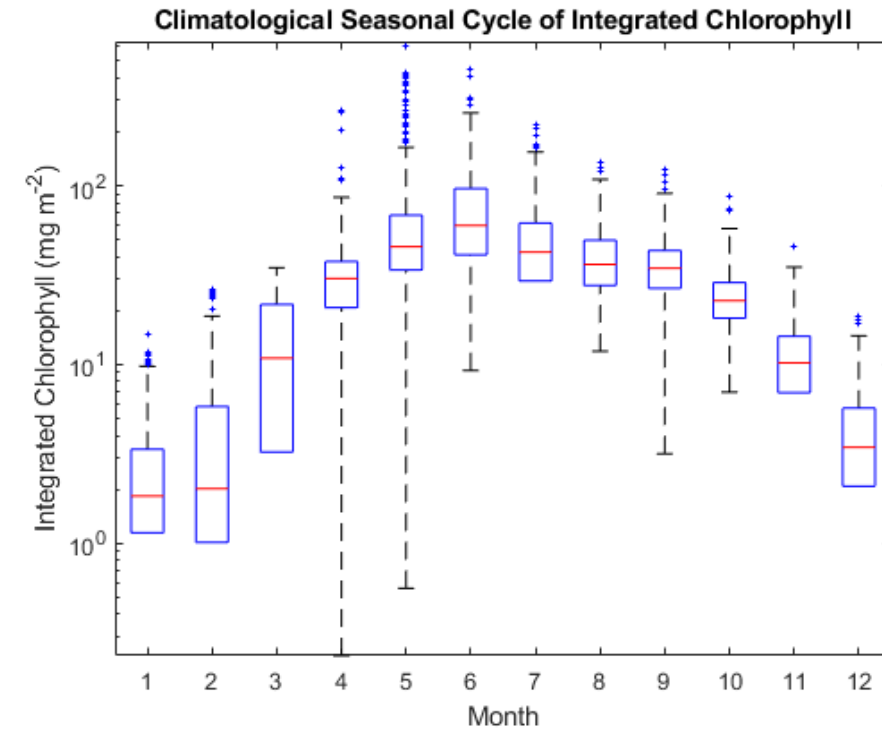
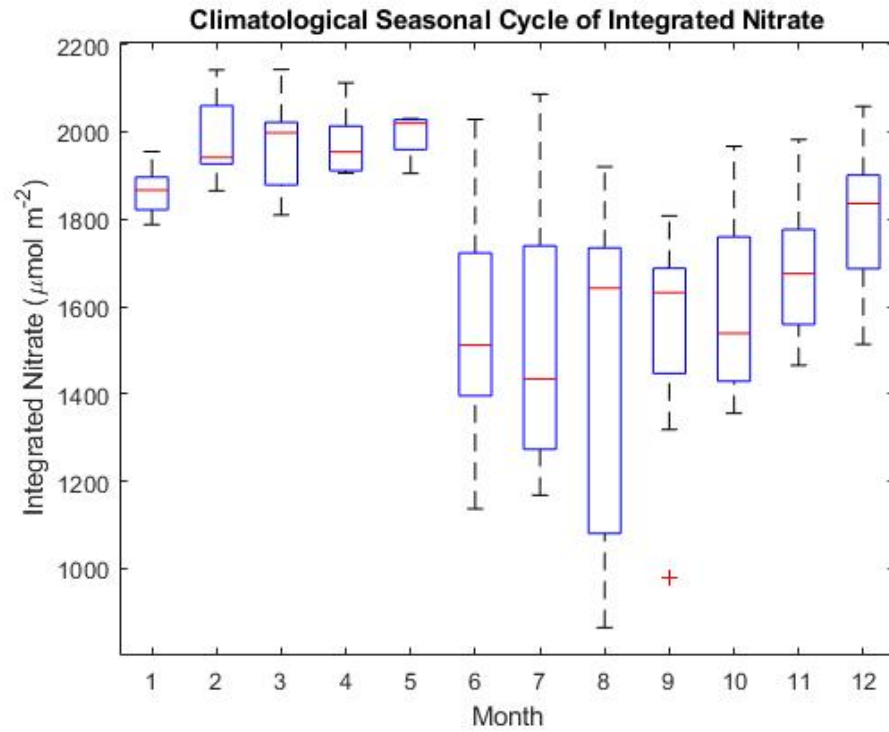
# Annual Net Community Production (ANCP) from Argo Floats

ANCP (circles) with Float Trajectories (lines)



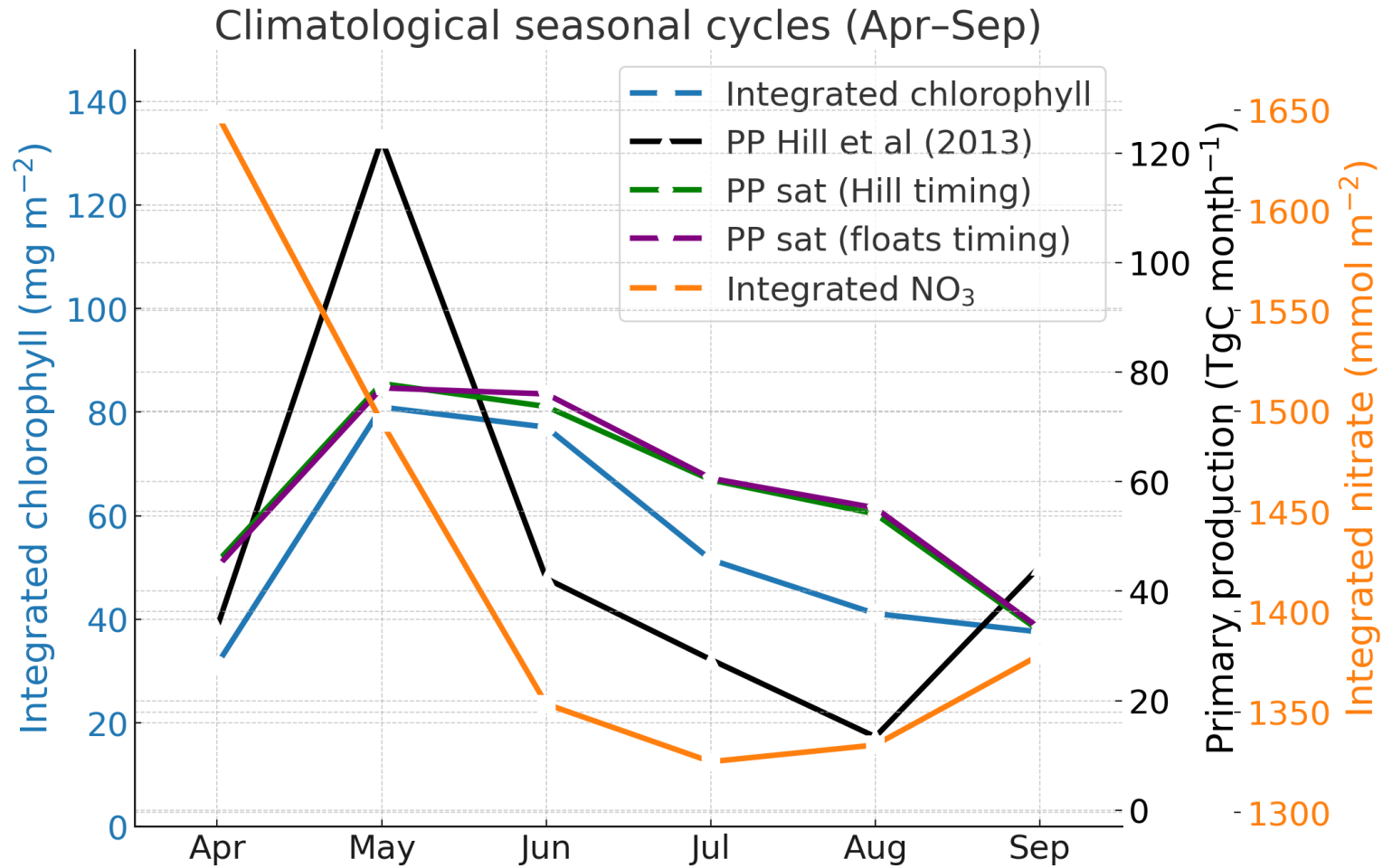
ANCP calculation method following Johnson et al. (2017)

# Climatological cycles





# BGC Argo CHL and satellite PP agree well



# Conclusions

PP algorithm is operational with Python codes uploaded online Accuracy of the selected model setups to reproduce the field data in terms of RMSD (RMSD=0.4) is better than in the related Arctic studies (RMSD=0.61-0.67)\*.

Larger Greenland Sea basin PP annual estimates, seasonal cycle pattern align with BGC Argo CHL

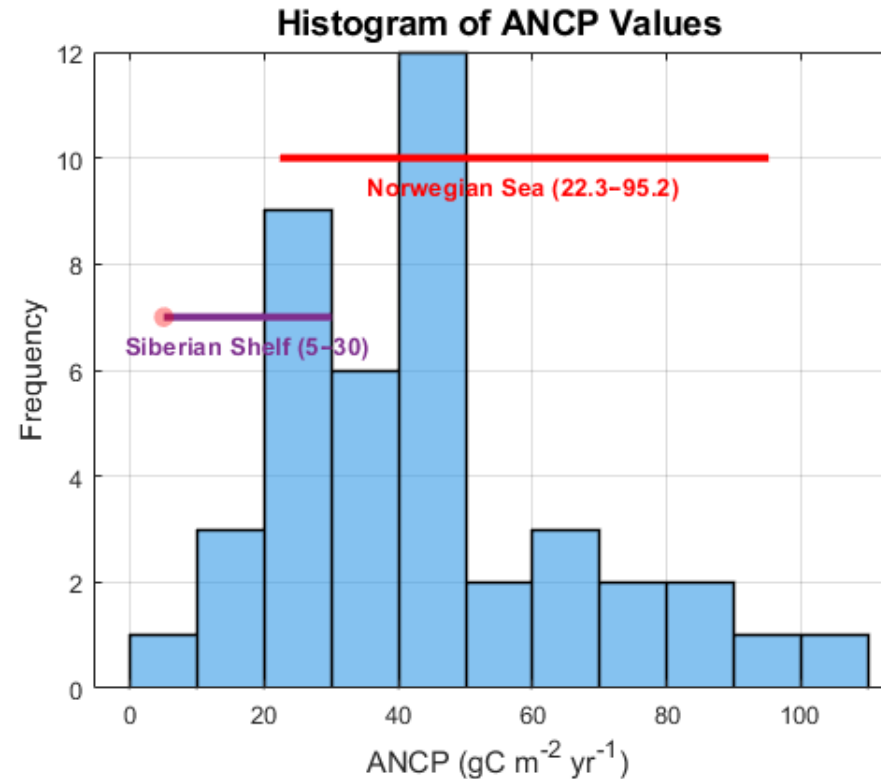
NCP algorithm is in progress as function of PP satellite, NCP BGC Argo estimates are in the range of glider studies

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Larger Greenland Sea

NCP algorithm is in pro



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C Argo estimates are in the

# Knowledge gaps and priorities for next steps



Scarce to no data for validation of PP and NCP at polar latitudes



Use not only traditional methods (C14), but also alternative (oxygen sensors, BGC floats)  
Include at least PP and maybe NCP as GOOS Essential Ocean Variable



Need for Arctic-specific CHL satellite algorithm



Zoffoli et al is planned to be soon available on Copernicus, better to use it if you're working in the Arctic



Need for reliable PAR level 3 product



Using climatologies or level 2 PAR from EUMETSAT instead



No salinity sensor for half of the BGC float dives (3445 out of 6881)



Include it in basic setup



# Global Observing Ocean System EOVS specification sheets

Physics	Biochemistry	Biology and Ecosystems
Sea state Ocean surface stress Sea ice Sea surface height Sea surface temperature Subsurface temperature Surface currents Subsurface currents Sea surface salinity Subsurface salinity Ocean surface heat flux Ocean bottom pressure Turbulent diapycnal fluxes (*pilot)	Oxygen Nutrients Inorganic carbon Transient tracers Particulate matter Nitrous oxide Stable carbon isotopes Dissolved organic carbon	Phytoplankton biomass and diversity Zooplankton biomass and diversity Fish abundance and distribution Sea turtles abundance and distribution Seabirds abundance and distribution Marine mammal abundance and distribution Coral cover and composition Seagrass cover and composition Macroalgal canopy cover and composition Mangrove cover and composition Microbe biomass and diversity (*pilot) Benthic invertebrate abundance and distribution (*pilot)
<b>Cross-disciplinary (including human impact)</b>		
	Ocean colour Marine debris (*pilot)	Ocean sound

## Phytoplankton Biomass and Diversity



# 1. EOVS information

## ESSENTIAL OCEAN VARIABLE (EOV)

Phytoplankton biomass and diversity

## DEFINITION

Phytoplankton biomass typically refers to either: weight (mass as the concentration per unit area/volume) and/or abundance or quantity of organisms (number of individuals per volume).

Phytoplankton diversity or composition refers to the variability among phytoplankton from all sources including, inter alia, marine and other aquatic ecosystems; this includes diversity within species and between species (e.g., genetic diversity, taxonomic diversity, size, etc.)

(<https://www.cbd.int/convention/articles?a=cbd-02> )

Cyanobacteria are primarily addressed here as key members of the phytoplankton, but also addressed in the microbes EOVS as members of the Bacteria.

## EOVS SUB-VARIABLES key measurements that are used to estimate the EOVS

\*Phytoplankton biomass (concentration)  
\*Composition (species, functional types)  
Number of HAB Events

\*bare minimum

## SUPPORTING VARIABLES - other measurements that are useful to provide scale or context to the sub-variables of the EOVS

Nutrients, sea surface temperature, subsurface temperature, sea surface salinity, subsurface salinity, oxygen, inorganic carbon, particulate organic matter concentration, total suspended organic matter concentration, ocean colour and bio-optical variables (remote sensing reflectance, absorption, scattering coefficients, photic or euphotic zone depth), mixed layer depth, surface currents and subsurface currents (vertical, horizontal), primary productivity

Complementary: cell size, "biovolume", nutritional content, density, DNA composition, ocean colour (bio-optics)

## DERIVED PRODUCTS outputs calculated from the EOVS and sub-variables, often in combination with the supporting variables

Phytoplankton Functional Types  
Diversity indices: species richness, species evenness, Simpson, etc.

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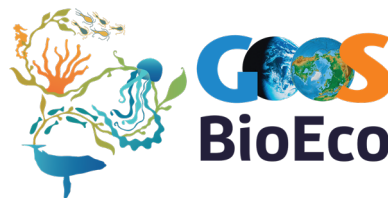
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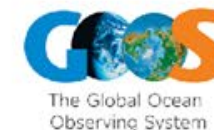




## Key stakeholders



## Partners



Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European Research Executive Agency (REA). Neither the European Union nor the granting authority can be held responsible for them.