







OceanBioME.jl: A flexible environment for modelling the coupled interactions between ocean biogeochemistry and physics

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Summary

OceanBioME.jl is a flexible modelling environment written in Julia ([Bezanson et al., 2017](#)) for modelling the coupled interactions between ocean biogeochemistry, carbonate chemistry, and physics. OceanBioME.jl can be used as a stand-alone box model, or integrated into Oceananigans.jl ([Ramadhan et al., 2020](#)) simulations of ocean dynamics in one, two, or three dimensions. As a result, OceanBioME.jl and Oceananigans.jl can be used to simulate the biogeochemical response across an enormous range of scales: from surface boundary layer turbulence at the meter scale to eddying global ocean simulations at the planetary scale, and on computational systems ranging from laptops to supercomputers. An example of a problem involving small-scale flow features is shown in [Figure 1](#), which shows a simulation of a sub-mesoscale eddy in a 1km × 1km horizontal domain with an intermediate complexity biogeochemical model and a kelp growth model solved along the trajectories of drifting buoys. OceanBioME.jl leverages Julia's multiple dispatch and effective inline capabilities to fuse its computations directly into existing Oceananigans.jl kernels, thus maintaining Oceananigans.jl's bespoke performance, memory- and cost-efficiency on GPUs in OceanBioME.jl-augmented simulations.

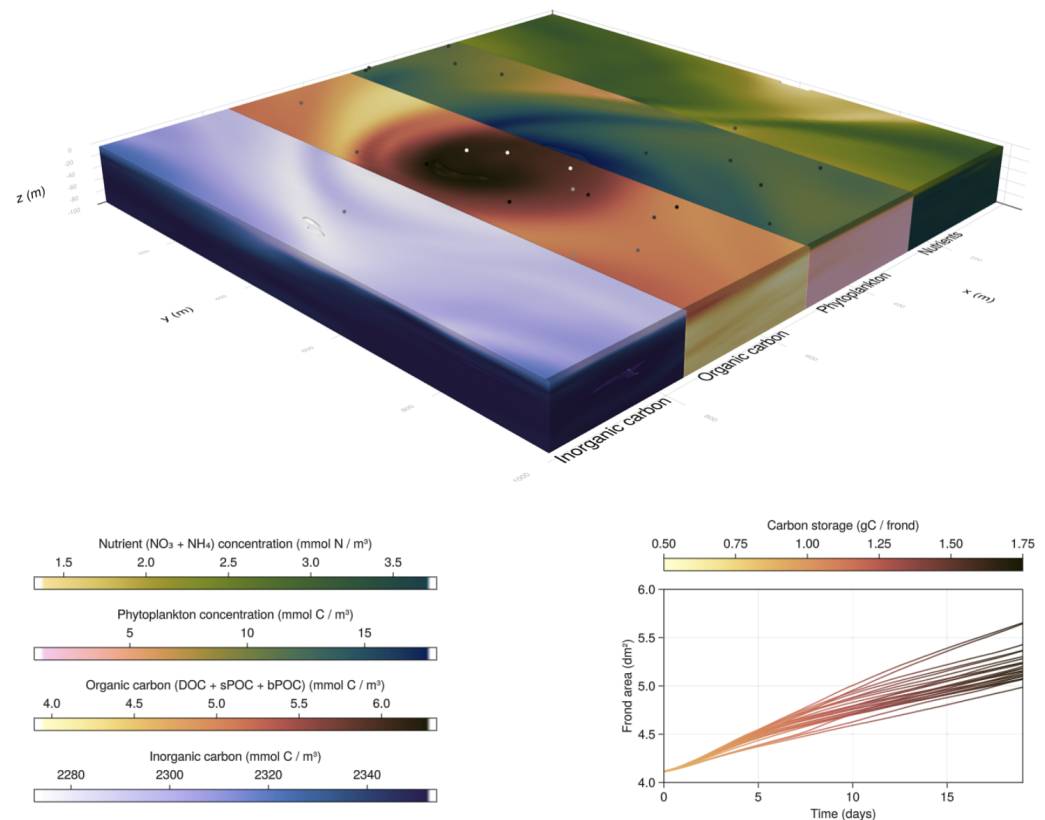


Figure 1: Here we replicate the Eady problem where a background buoyancy gradient and corresponding thermal wind generate a sub-mesoscale eddy, roughly following the setup of Taylor (2016). To this physical setup, we added a medium complexity (9 tracers) biogeochemical model, some of which are shown above. On top of this, we added particles modelling the growth of sugar kelp which are free-floating and advected by the flow, and carbon dioxide exchange from the air. Thanks to Julia's speed and efficiency the above model ($1 \text{ km} \times 1 \text{ km} \times 100 \text{ m}$ with $64 \times 64 \times 16$ grid points) took about 30 minutes of computing time to simulate 10 days of evolution on an Nvidia P100 GPU. Figure made with Makie.jl (Danisch & Krumbiegel, 2021).

26 OceanBioME.jl is built with a highly modular design that allows user control and customization.
27 There are two distinct module types implemented in OceanBioME.jl:

- 28 ■ First, tracer-based ecosystem modules are formulated in AdvectedPopulations as a set of
29 coupled ordinary differential equations. These equations can be solved by OceanBioME.jl
30 as box models, which is particularly useful for testing. The same equations can be
31 integrated by Oceananigans.jl to provide tracer-based ecosystem models.
- 32 ■ The second module type is Individual “biologically active” particles. These consist
33 of individual-based models which are solved along particle paths and can be coupled
34 with the tracer-based modules and physics from Oceananigans.jl. The biologically
35 active particles can be advected by the currents, and/or they can move according to
36 prescribed dynamics. For example, migrating zooplankton or fish can be modelled with
37 biologically active particles and OceanBioME.jl allows these to interact with tracer-based
38 components such as phytoplankton or oxygen.

39 AdvectedPopulations are supported by Boundaries modules which provide information at
40 the top and bottom of the ocean. For example, the GasExchange submodule calculates the
41 flux of carbon dioxide and oxygen at the sea surface, while the Sediments modules calculates
42 fluxes of carbon and oxygen at the seafloor.

43 We provide a simple framework and utilities (such as light attenuation integration) to build the
44 necessary components of biogeochemical models. With the provided models, currently a simple
45 Nutrient-Phytoplankton-Zooplankton-Detritus (Kuhn et al., 2015) model, and an intermediate
46 complexity model, LOBSTER (Lévy et al., 2005), we have set up a straightforward “plug and
47 play” framework to add additional tracers such as carbonate and oxygen chemistry systems,
48 and additional forcing. Additionally, we have implemented comprehensive air-sea flux models
49 (e.g. Wanninkhof, 1992) and sediment models (e.g. Soetaert et al., 2000) which can easily be
50 applied to tracers in the models. We focus on the simulation of idealized sub-mesoscale systems,
51 but this flexible framework allows users to model problems of any scale. This framework is
52 made possible by our contributions to Oceananigans.jl, adding a streamlined user interface
53 to swap biogeochemical models with no modification to other model configurations. This
54 interface also facilitates rapid prototyping, as models can be implemented and swapped easily
55 by just extending a few key functions. This flexibility and ease-of-use is unmatched in existing
56 biogeochemical models.

57 OceanBioME.jl was designed specifically to study ocean carbon dioxide removal (OCDR)
58 strategies. Assessing the effectiveness and impacts of OCDR is challenging due to the
59 complexities of the interactions between the biological, chemical, and physical processes
60 involved in the carbon cycle. Moreover, field trials of OCDR interventions are generally small-
61 scale and targeted, while the intervention required to have a climate-scale impact is regional or
62 global. We have built OceanBioME.jl to meet these challenges by creating tools that provide a
63 modular interface to the different components within the ocean modelling framework provided
64 by Oceananigans.jl. This allows easy access to a suite of biogeochemical models ranging
65 from simple idealized to full-complexity models. Figure 2 shows a simple column model with
66 an OCDR intervention (macroalgae growth) added after a warm-up period, which increases
67 the carbon export of the system.

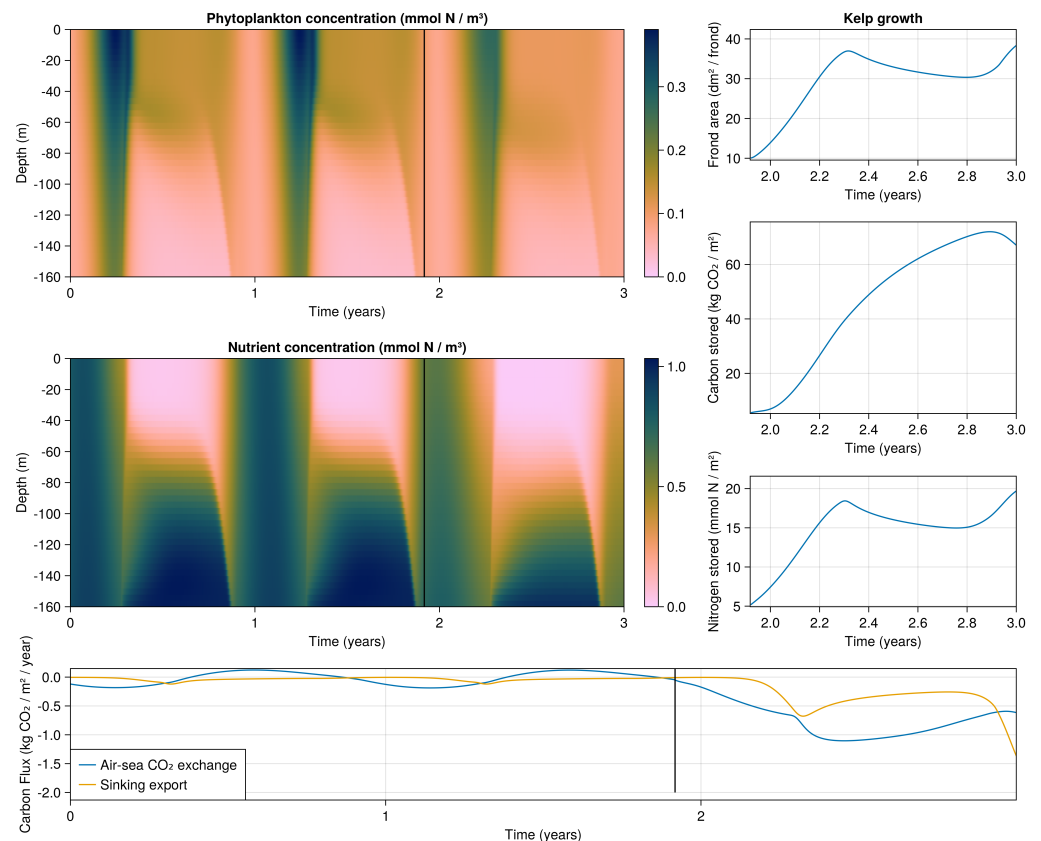


Figure 2: Here we show the results of a 1D model, forced by idealised light and mixing, which qualitatively reproduces the biogeochemical cycles in the North Atlantic. We then add kelp (500 frond / m² in the top 50 m of water) in December of the 2nd year (black vertical line) which causes an increase in air-sea carbon dioxide exchange and sinking export, as well as a change in the phytoplankton growth cycle. Figure made with Makie.jl (Danisch & Krumbiegel, 2021).

68 The biologically active particles built into OceanBioME.jl are particularly useful for OCDR
69 applications. Accurate carbon accounting is essential for assessing the effectiveness of OCDR
70 strategies. Biologically active particles can be used to track carbon from a particular source while
71 accounting for interactions with its surroundings. Biologically active particles can also be used
72 to model OCDR deployment strategies including seaweed cultivation, alkalinity enhancement,
73 and marine biomass regeneration. OceanBioME.jl currently includes an extended version of
74 the sugar kelp model presented by Broch & Slagstad (2012) as an example of the utility and
75 implementation of these features.

76 The implementation of OceanBioME.jl models allows for seamless integration with data
77 assimilation packages, such as EnsembleKalmanProcesses.jl (Dunbar et al., 2022). This
78 feature facilitates rapid calibration of model parameters, providing a powerful utility for
79 integrating observations and models, with the potential to improve model skill and identify key
80 sources of uncertainty.

81 A key metric for the validity of biogeochemical systems is the conservation of elements such as
82 carbon and nitrogen in the system. We therefore continuously test the implemented models in
83 a variety of simple scenarios (i.e. isolated, with/without air-sea flux, with/without sediment)
84 to ensure that conservation conditions are met, and we will continue to add tests for any
85 new models. Additionally, we check OceanBioME.jl utilities through standard tests such as
86 comparison to analytical solutions for light attenuation, and conservation of tracers for active
87 particle exudation and sinking.

88 Finally, this software is currently facilitating multiple research projects into ocean CDR which
 89 would have been significantly harder with other solutions. For example, Chen (In prep.) is
 90 using the active particle coupling provided to investigate the effects of location and planting
 91 density of kelp in the open ocean on their carbon drawdown effect, as in the example above.
 92 Additionally, Strong-Wright (In prep.) is using the coupling of both the biogeochemistry and
 93 easy interface to couple the physics to study flow interactions with a fully resolved giant kelp
 94 forest model including the effects on nutrient transport and distribution.

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