




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

Jago Strong-Wright^{1,2}✉, Si Chen^{1,2}, Navid C Constantinou^{3,4}, Simone Silvestri⁵, Gregory LeClaire Wagner⁵, and John R Taylor^{1,2}

¹ Department of Applied Mathematics and Theoretical Physics, University of Cambridge, Cambridge, United Kingdom ² Centre for Climate Repair, Cambridge, United Kingdom ³ Australian National University, Australia ⁴ Australian Research Council Centre for Climate Extremes ⁵ Massachusetts Institute of Technology, USA ✉ Corresponding author

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

Software

- [Review](#) 
- [Repository](#) 
- [Archive](#) 

Editor: [Kristen Thyng](#) 

Reviewers:

- [@francispoulin](#)
- [@iuryt](#)

Submitted: 24 June 2023

Published: unpublished

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC BY 4.0](https://creativecommons.org/licenses/by/4.0/))

Statement of Need

To date, about 25% of anthropogenic carbon emissions have been taken up by the ocean (Friedlingstein et al., 2022). This occurs through complex interactions between physics, chemistry, and biology, much of which is poorly understood. Due to the vast size of the ocean and the sparsity of data, modelling and data assimilation play a vital role in quantifying the ocean carbon cycle. Traditionally ocean biogeochemical (BGC) modelling involves large and inflexible code bases written in high-performance but low-level languages which require huge computational resources to execute. This causes a barrier to experimentation and innovation as users must develop expertise in both the science and complex code.

One area where novel ideas must be explored with BGC codes is assessing ocean carbon dioxide removal (OCDR) strategies. Quantifying the effectiveness and identifying the impacts of OCDR is challenging due to the aforementioned complexity of the ocean BGC system. Moreover, field trials of OCDR interventions are generally small-scale and targeted, while the intervention required to have a climate-scale impact is regional or global. This necessitates adaptable, easy-to-use, and verifiable BGC modelling tools which can be used to assess OCDR strategies at the fast pace with which they are being developed (National Academies of Sciences & Medicine, 2022). We have built OceanBioME.jl to meet these challenges by creating a tool that provides a modular interface to the different components, within the ocean modelling framework provided by Oceananigans.jl. This allows easy access to a suite of biogeochemical models ranging from simple idealized to full-complexity models. The flexibility of the Oceananigans.jl framework allows OceanBioME.jl to be applied across a wide range of scales and use cases, including small-scale large-eddy simulations and regional and global models.

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 x 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.

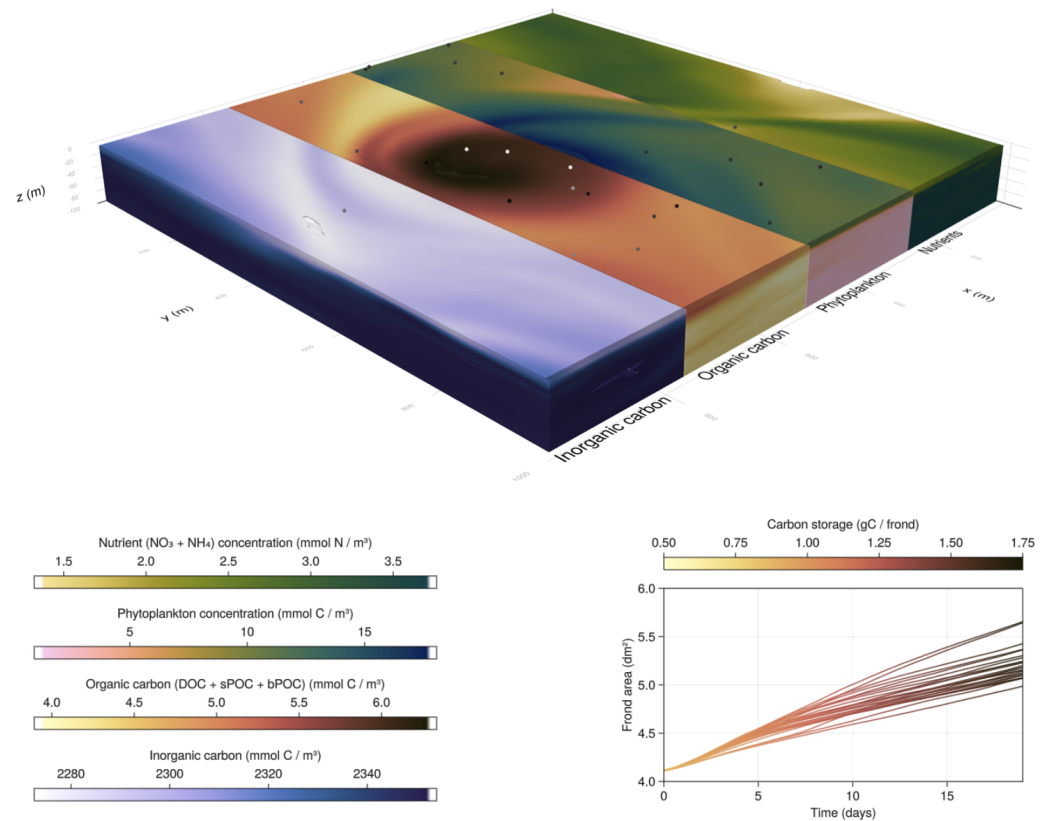


Figure 1: Here we replicate the Eady problem where a background buoyancy gradient and corresponding thermal wind generate a sub-mesoscale eddy, roughly following the setup of Taylor (2016). To this physical setup, we added a medium complexity (9 tracers) biogeochemical model, some of which are shown 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 km × 1 km × 100 m with 64 × 64 × 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).

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

- First, tracer-based ecosystem modules are formulated in AdvectedPopulations as a set of coupled ordinary differential equations. These equations can be solved by OceanBioME.jl as box models, which is particularly useful for testing. The same equations can be integrated by Oceananigans.jl to provide tracer-based ecosystem models.
- The second module type is Individual "biologically active" particles. These consist of individual-based models solved along particle paths and can be coupled with the tracer-based modules and physics from Oceananigans.jl. The biologically active particles can be advected by the currents, and/or they can move according to prescribed dynamics.

For example, migrating zooplankton or fish can be modelled with biologically active particles and OceanBioME.jl allows these to interact with tracer-based components such as phytoplankton or oxygen.

For example, the GasExchange submodule calculates the carbon dioxide and oxygen flux at the sea surface, while the Sediments modules calculate fluxes of carbon and oxygen at the seafloor.

We currently provide a simple Nutrient-Phytoplankton-Zooplankton-Detritus (NPZD) model (Kuhn et al., 2015), and an intermediate complexity model, LOBSTER (Lévy et al., 2005), we have set up a straightforward “plug and play” framework to add additional tracers such as carbonate and oxygen chemistry systems and additional forcing. These AdvectedPopulations are supported by Boundaries modules which are easy to apply and provide information at the top and bottom of the ocean. We have implemented comprehensive air-sea flux models (e.g. Wanninkhof, 1992) within the GasExchange submodule to calculate carbon dioxide and oxygen flux at the sea surface, and sediment models (e.g. Soetaert et al., 2000) which calculate fluxes of carbon and oxygen at the seafloor. We focus on the simulation of idealized sub-mesoscale systems, but this flexible framework allows users to model problems of any scale. For example, Figure 2 shows the annual average surface phytoplankton concentration from a near-global model NPZD model run. This framework is made possible by our contributions to Oceananigans.jl, adding a streamlined user interface to swap biogeochemical models with no modification to other model configurations. This interface also facilitates rapid prototyping, as models can be implemented and swapped easily by just extending a few key functions. This flexibility and ease-of-use is unmatched in existing biogeochemical models.

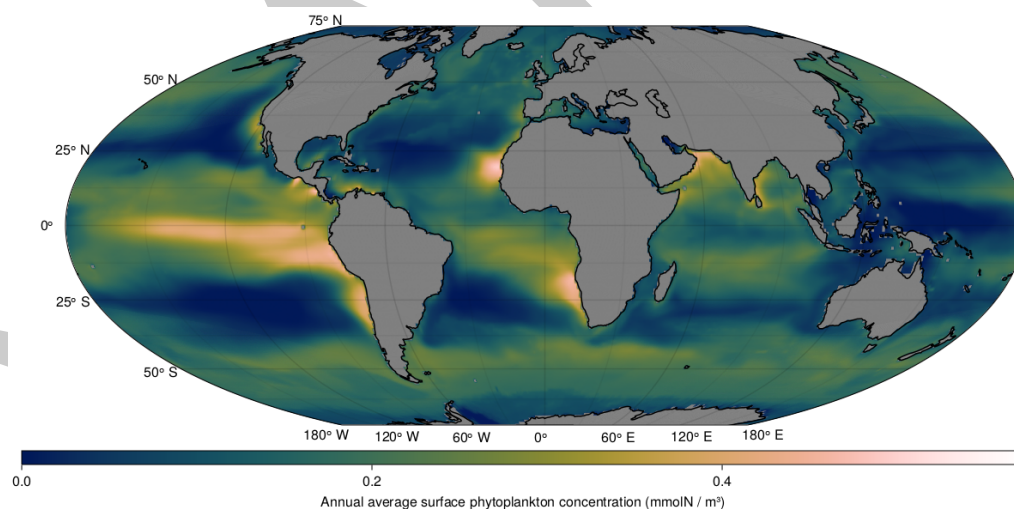


Figure 2: Here we show the annual average surface phytoplankton concentration from a near-global NPZD model run. It shows reasonably good reproduction of large-scale patterns for such a simple and uncalibrated model but demonstrates further work such as nutrient input from rivers and tuning physics parametrisations that are required in the future. We ran this model with a 1° horizontal resolution and 48 (irregularly spaced) vertical points, and it took around 45 minutes per year to run on an Nvidia A100 GPU. Figure made with Makie.jl (Danisch & Krumbiegel, 2021).

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

88 implementation of these features. Figure 3 shows a simple column model with an OCDR
89 intervention (macroalgae growth) added after a warm-up period, which increases the carbon
90 export of the system.

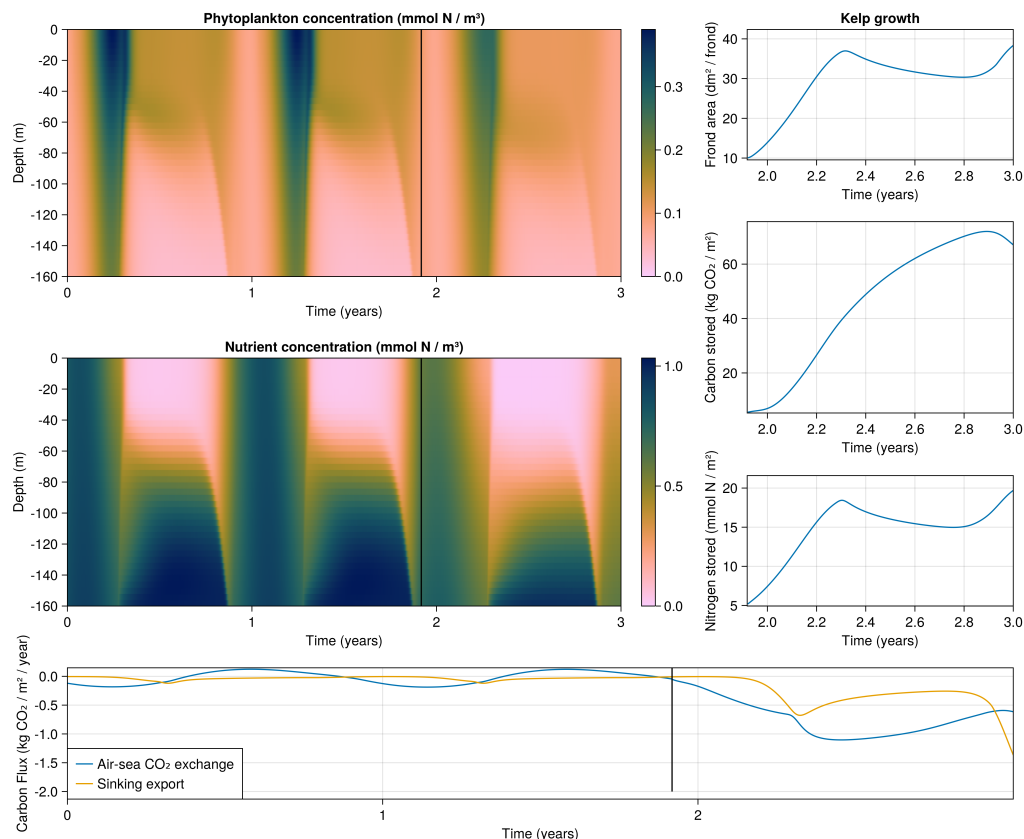


Figure 3: Here we show the results of a 1D model, forced by idealised light and mixing, which qualitatively reproduces the biogeochemical cycles in the North Atlantic. We then add kelp (500 frond / m² in the top 50 m of water) in December of the 2nd year (black vertical line) which causes an increase in air-sea carbon dioxide exchange and sinking export. Changes to the phytoplankton growth cycle are also apparent. Figure made with Makie.jl (Danisch & Krumbiegel, 2021).

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

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

103 Finally, this software is currently facilitating multiple research projects into ocean CDR which
104 would have been significantly harder with other solutions. For example, Chen (In prep.) is
105 using the active particle coupling provided to investigate the effects of location and planting
106 density of kelp in the open ocean on their carbon drawdown effect, as in the example above.

107 Additionally, Strong-Wright (In prep.) is using the coupling of both the biogeochemistry and
 108 easy interface to couple the physics to study flow interactions with a fully resolved giant kelp
 109 forest model including the effects on nutrient transport and distribution.

110 Acknowledgements

111 We would like to thank the [Climate Modeling Alliance](#) team and `Oceananigans.jl` contributors
 112 for their fantastic project. We are also very grateful for the support and funding of the [Centre](#)
 113 [for Climate Repair, Cambridge](#) and the [Gordon and Betty Moore Foundation](#).

114 References

- 115 Bezanson, J., Edelman, A., Karpinski, S., & Shah, V. B. (2017). Julia: A fresh approach to
 116 numerical computing. *SIAM Review*, 59(1), 65–98. <https://doi.org/10.1137/141000671>
- 117 Broch, O. J., & Slagstad, D. (2012). Modelling seasonal growth and composition of the kelp
 118 *saccharina latissima*. *Journal of Applied Phycology*, 24, 759–776. <https://doi.org/10.1007/s10811-011-9695-y>
- 119 Danisch, S., & Krumbiegel, J. (2021). Makie.jl: Flexible high-performance data visualization for
 120 julia. *Journal of Open Source Software*, 6(65), 3349. <https://doi.org/10.21105/joss.03349>
- 121 Dunbar, O. R. a., Lopez-Gomez, I., Garbuno-Iñigo, A., Huang, D. Z., Bach, E., & Wu, J.
 122 (2022). EnsembleKalmanProcesses.jl: Derivative-free ensemble-based model calibration.
 123 *Journal of Open Source Software*, 7(80), 4869. <https://doi.org/10.21105/joss.04869>
- 124 Friedlingstein, P., Jones, M. W., O'Sullivan, M., Andrew, R. M., Bakker, D. C. E., Hauck, J.,
 125 Le Quéré, C., Peters, G. P., Peters, W., Pongratz, J., Sitch, S., Canadell, J. G., Ciais, P.,
 126 Jackson, R. B., Alin, S. R., Anthoni, P., Bates, N. R., Becker, M., Bellouin, N., ... Zeng,
 127 J. (2022). Global Carbon Budget 2021. *Earth System Science Data*, 14(4), 1917–2005.
 128 <https://doi.org/10.5194/essd-14-1917-2022>
- 129 Kuhn, A. M., Fennel, K., & Mattern, J. P. (2015). Model investigations of the north atlantic
 130 spring bloom initiation. *Progress in Oceanography*, 138, 176–193. <https://doi.org/10.1016/j.pocean.2015.07.004>
- 131 Lévy, M., Gavart, M., Mémary, L., Caniaux, G., & Paci, A. (2005). A four-dimensional
 132 mesoscale map of the spring bloom in the northeast atlantic (POMME experiment):
 133 Results of a prognostic model. *Journal of Geophysical Research C: Oceans*, 110, 1–23.
 134 <https://doi.org/10.1029/2004JC002588>
- 135 National Academies of Sciences, Engineering, & Medicine. (2022). *A research strategy for*
 136 *ocean-based carbon dioxide removal and sequestration*. The National Academies Press.
 137 <https://doi.org/10.17226/26278>
- 138 Ramadhan, A., Wagner, G. L., Hill, C., Jean-Michel, C., Churavy, V., Souza, A., Edelman,
 139 A., Ferrari, R., & Marshall, J. (2020). Oceananigans.jl: Fast and friendly geophysical fluid
 140 dynamics on GPUs. *Journal of Open Source Software*, 5(53), 2018. <https://doi.org/10.21105/joss.02018>
- 141 Soetaert, K., Middelburg, J. J., Herman, P. M. J., & Buis, K. (2000). On the coupling
 142 of benthic and pelagic biogeochemical models. In *Earth-Science Reviews* (Vol. 51, pp.
 143 173–201). [https://doi.org/10.1016/S0012-8252\(00\)00004-0](https://doi.org/10.1016/S0012-8252(00)00004-0)
- 144 Wanninkhof, R. (1992). Relationship between wind speed and gas exchange over the ocean.
 145 *Journal of Geophysical Research*, 97, 7373–7382. <https://doi.org/10.1029/92JC00188>