

¹ ar-bgc-argo: Jupyter Notebook templates for
² searching, downloading, and post-processing
³ biogeochemical Argo float time series

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

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Submitted: 01 January 1970

Published: unpublished

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Summary

ar-bgc-argo is a set of Jupyter Notebook templates that transform raw profiles of biogeochemical Argo (BGC-Argo) floats into “analysis-ready” time series of ocean temperature, salinity, and biogeochemical variables. Users can search for floats based the dates, geographic region, and biogeochemical variables of their interests. After downloading the profile time series of a selected float, ar-bgc-argo can visualize, filter, interpolate, and save the post-processed time series as a netCDF file. In addition, ar-bgc-argo applies variable-specific treatments and derive additional oceanographic variables using empirical equations. ar-bgc-argo is designed to help expand the end users of the growing BGC-Argo float data.

Statement of need

Biogeochemical Argo (BGC-Argo) is a global network of autonomous profiling floats in the ocean that has brought dramatic advances in our understanding of ocean biogeochemistry and marine ecosystems in recent years ([Thierry et al., 2025](#)). The global BGC-Argo community aims to cover the global ocean with 1,000 floats that operate every 10 days or so, monitoring biogeochemical properties from the sea surface to 2,000 m deep ([Claustre et al., 2020](#)). These profiling floats are equipped with sensors that can measure up to six key variables: chlorophyll-a, pH, oxygen, nitrate, irradiance, and suspended particles ([Bittig et al., 2019](#)). As of December 2025, 861 BGC-Argo floats are already in operation, but only 111 of these are equipped with the six full sensors (Figure 1). The collected profiles are made publicly available in the netCDF format within a day or so ([Wong et al., 2020](#)).

Despite the growing application of BGC-Argo data, the raw profiles include technical errors and doubtful values because of poor sensor calibration and high sensitivity to noise and artifacts. Furthermore, the data are unfiltered (containing both good- and bad-quality samples) and have inconsistent sampling depths among profiles from a given float. These issues necessitate post-processing prior to scientific analysis, which requires technical knowledge hence, becomes a time-consuming task.

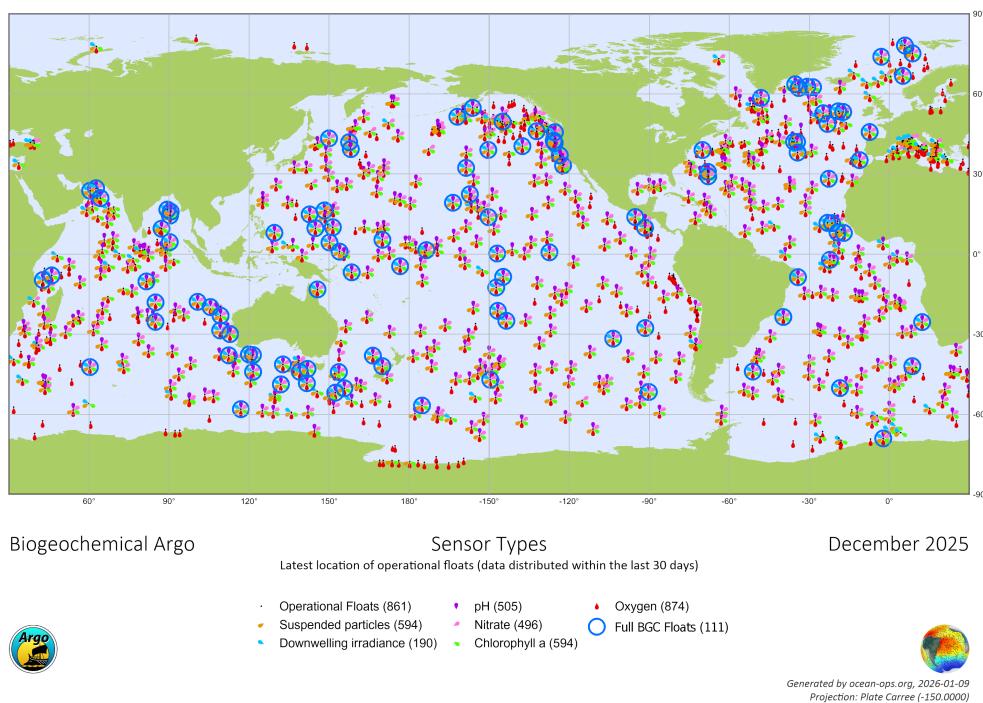


Figure 1: Figure 1: Global coverage of operational BGC-Argo floats as of December 2025 (<https://www.ocean-ops.org/share/Argo/Maps/bgc.png>; accessed on January 16, 2026).

34 Overview of ar-bgc-argo

35 ar-bgc-argo consists of three Jupyter Notebook templates: search.ipynb; download.ipynb;
 36 and generate.ipynb. To use these templates, users create a copy of the template of their
 37 interest, modify its inputs, and run it on their Jupyter environment.

38 search.ipynb

39 search.ipynb searches for BGC-Argo floats from the synthetic-profile index file (argo_synthetic-
 40 profile_index.txt) of the Global Data Assembly Center (GDAC; (Bittig et al., 2019)) and
 41 based on the user inputs, including the temporal and spatial coverages and the biogeochemical
 42 variables of interest. In addition, search.ipynb allows users to narrow down the float selection
 43 based on three key criteria:

- 44 ■ mindays: the minimum duration of the data record to ensure sufficient temporal coverage
 (e.g., at least 365 days).
- 45 ■ minfreq: the minimum sampling frequency required to capture temporal variability (e.g.,
 at least every 14 days).
- 46 ■ maxdrift: the float's maximum drift speed (e.g., 0.05 m/s), which is particularly use-
 ful for identifying "quasi-Eulerian" floats suitable for one-dimensional modeling (e.g.,
 (Bruggeman et al., 2024)).

51 The trajectories of all qualified floats based on the search criteria are drawn on a map and their
 52 temporal coverages are visualized on a time series. These visualizations provide an intuitive
 53 overview, enabling users to identify potential spatial and temporal biases in observational
 54 coverage within the study region of interest prior to data retrieval (Hayashida et al., 2025).

55 download.ipynb

56 download.ipynb retrieves the concatenated synthetic-profiles time series of a selected float
 57 ([wmoid]_Sprof.nc, where [wmoid] is the seven-digit World Meteorological Organization Identifier or WMO ID). While download.ipynb naturally follows the selection made in search.ipynb,
 58 it can also be used independently if the WMO ID of the target float is already known. Upon
 59 execution, download.ipynb creates a directory named after the WMO ID to store the data.
 60 It then identifies the correct file path from the synthetic-profile index file and downloads the
 61 data from one of the two GDACs.

63 generate.ipynb

64 generate.ipynb is the core component of ar-bgc-argo, designed to transform raw BGC-Argo
 65 profiles into “analysis-ready” time series suitable for immediate scientific application. The
 66 data processing workflow consists of eight steps (Figure 2). At every step, diagnostic plots
 67 are generated and enable users to visually verify the reliability of the post-processing, which
 68 prevents a “black-box” approach and ensuring the production of high-quality datasets.

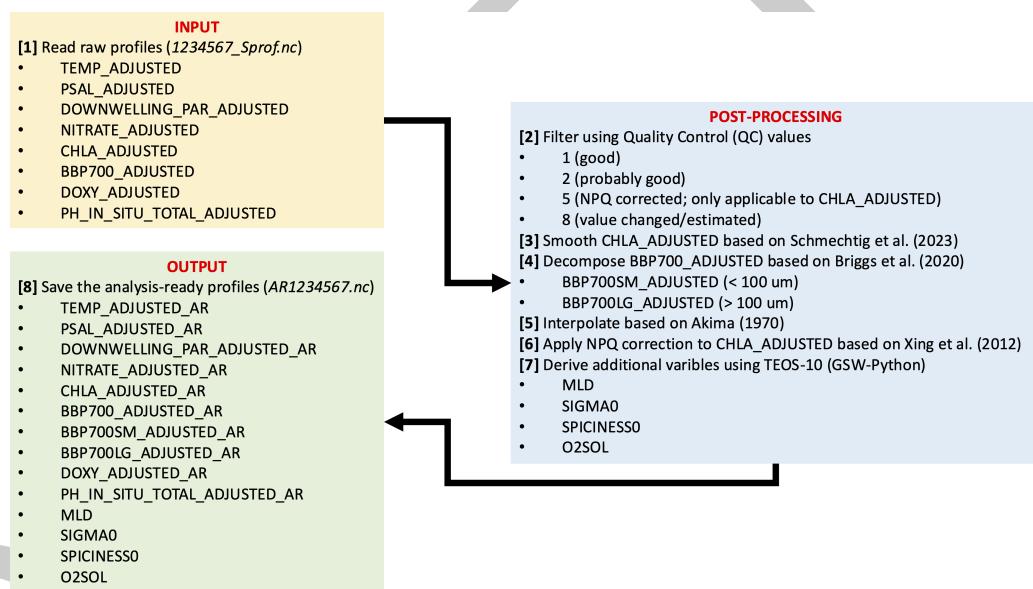


Figure 2: Figure 2: Schematic workflow of the data-processing pipeline implemented in generate.ipynb, which involves procedures based on Briggs et al. (2020), Schmechtig et al. (2023), and Xing et al. (2012). This example assumes a full-sensor float with an arbitrary WMO ID (1234567).

69 Acknowledgements

70 BGC-Argo data are freely available through one of the two GDACs. We thank Hidehiro Fujio
 71 for computational assistance and Pete Strutton, Clara Vives, Kanako Sato, Yoshimi Kawai
 72 for insightful discussion on BGC-Argo floats and feedback on earlier versions of ar-bgc-argo.
 73 This work was supported by JSPS KAKENHI Grant Number JP24H02226 and the Nakajima
 74 Foundation.

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