

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

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


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8 Summary

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

17 Statement of need

18 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
22 biogeochemical properties from the sea surface to 2,000 m deep ([Claustre et al., 2020](#)). These
23 profiling floats are equipped with sensors that can measure up to six key variables: chlorophyll-a,
24 pH, oxygen, nitrate, irradiance, and suspended particles ([Bittig et al., 2019](#)). As of December
25 2025, 861 BGC-Argo floats are already in operation, but only 111 of these are equipped with
26 the six full sensors (Figure 1). The collected profiles are made publicly available in the netCDF
27 format within a day or so ([Wong et al., 2020](#)).

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

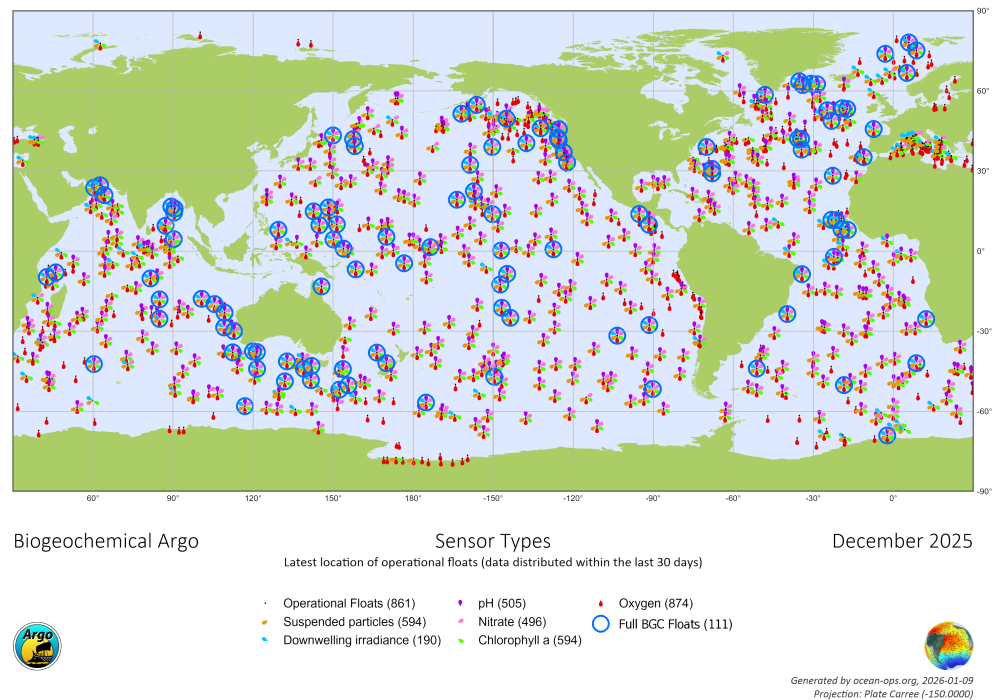


Figure 1: Figure 1: Global coverage of operational BGC-Argo floats as of December 2025 (<https://www.ocean-ops.org/share/Argo/Maps/bgc.png>; accessed on January 16, 2026).

Overview of ar-bgc-argo

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

search.ipynb

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

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

The trajectories of all qualified floats based on the search criteria are drawn on a map and their temporal coverages are visualized on a time series. These visualizations provide an intuitive overview, enabling users to identify potential spatial and temporal biases in observational 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.
62

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.

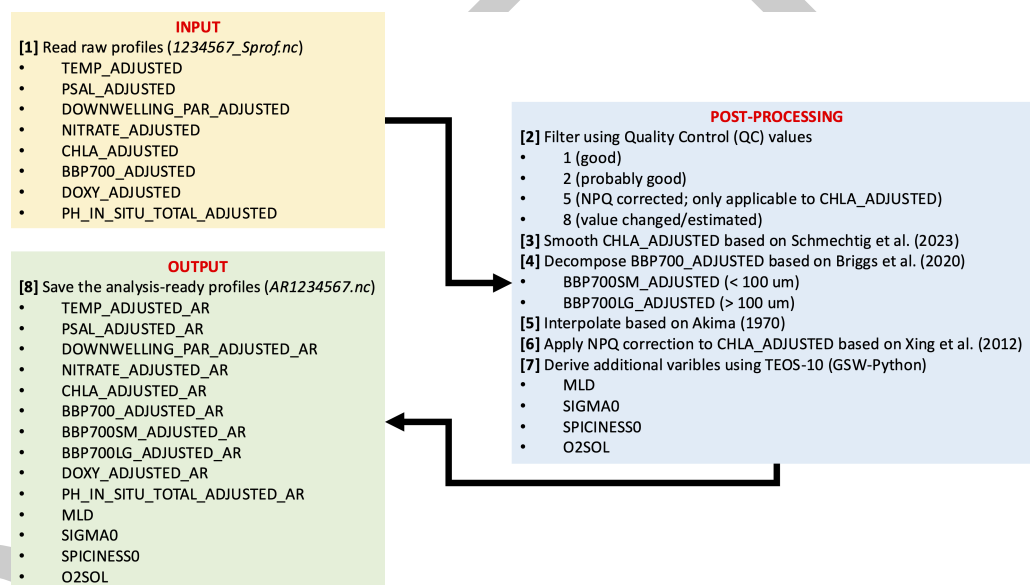


Figure 2: Figure 2: Schematic workflow of the data-processing pipeline implemented in generate.ipynb, which involves procedures based on Briggs et al. (2020), Schmechtig et al. (2023), and Xing et al. (2012). This example assumes a full-sensor float with an arbitrary WMO ID (1234567).

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