

ROMS-Tools: Reproducible, Scalable Preprocessing and Analysis for Regional Ocean Modeling with ROMS

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Summary

The ocean regulates Earth's climate and sustains marine ecosystems by circulating and storing heat, carbon, oxygen, and nutrients, while exchanging heat and gases with the atmosphere. Scientists study these processes using ocean models, which simulate the ocean on a grid. **Regional ocean models** focus computational resources on a limited geographical area with fine grid spacing, and can resolve fine-scale phenomena such as mesoscale and submesoscale features, tidal dynamics, coastal currents, upwelling, and detailed biogeochemical (BGC) processes. A widely used regional ocean model is the **Regional Ocean Modeling System (ROMS)** (Shchepetkin & McWilliams, 2005). ROMS has been coupled to the Marine Biogeochemistry Library (MARBL) (Long et al., 2021; Molemaker & contributors, 2025a) to link physical and BGC processes. ROMS-MARBL supports research on environmental management, fisheries, regional climate impacts, and ocean-based carbon dioxide removal (CDR) strategies.

Configuring a regional ocean model like ROMS-MARBL is technically challenging; it requires initialization and time-dependent forcing from oceanic and atmospheric data drawn from multiple external sources in diverse formats. These global datasets can reach several petabytes and must be subsetting, processed, and mapped onto the target domain's geometry, yielding input datasets of 10–100 terabytes for large regional models. Generating these input files is time-consuming, error-prone, and hard to reproduce, creating a bottleneck for both new and experienced users. The Python package ROMS-Tools addresses this challenge by providing efficient, dask-backed (Dask Development Team, 2016), user-friendly tools that can be installed via Conda or PyPI and run interactively from Jupyter notebooks. It supports creating regional grids, preprocessing all required model inputs, and postprocessing and analysis. Current capabilities are fully compatible with UCLA-ROMS (Molemaker & contributors, 2025a, 2025b), with potential support for other ROMS versions, such as Rutgers ROMS (Arango & contributors, 2024), in the future.

Input Data and Preprocessing

ROMS-Tools generates the following input files for ROMS-MARBL:

- Model Grid:** Customizable, curvilinear, and orthogonal grid designed to maintain a nearly uniform horizontal resolution across the domain. The grid is rotatable to align with coastlines and features a terrain-following vertical coordinate.
- Bathymetry:** Derived from **SRTM15** (Tozer et al., 2019).
- Land Mask:** Inferred from coastlines provided by **Natural Earth** or the Global Self-consistent, Hierarchical, High-resolution Geography (**GSHHG**) Database (Vessel &

- Smith, 1996).
4. **Physical Ocean Conditions:** Initial and open boundary conditions for sea surface height, temperature, salinity, and velocities derived from the 1/12° Global Ocean Physics Reanalysis (GLORYS) (Lellouche et al., 2021).
 5. **BGC Ocean Conditions:** Initial and open boundary conditions for dissolved inorganic carbon, alkalinity, and other biogeochemical tracers from Community Earth System Model (CESM) output (Yeager et al., 2022) or hybrid observational-model sources (Garcia et al., 2019; Huang et al., 2022; Lauvset et al., 2016; Yang et al., 2020; Yeager et al., 2022).
 6. **Meteorological forcing:** Wind, radiation, precipitation, and air temperature/humidity processed from the global 1/4° ECMWF Reanalysis v5 (ERA5) (Hersbach et al., 2020) with optional corrections for radiation bias and coastal wind.
 7. **BGC surface forcing:** Partial pressure of carbon dioxide, as well as iron, dust, and nitrogen deposition from CESM output (Yeager et al., 2022) or hybrid observational-model sources (Hamilton et al., 2022; Kok et al., 2021; Landschützer et al., 2016; Yeager et al., 2022).
 8. **Tidal Forcing:** Tidal potential, elevation, and velocities derived from TPXO (Egbert & Erofeeva, 2002) including self-attraction and loading (SAL) corrections.
 9. **River Forcing:** Freshwater runoff derived from Dai & Trenberth (Dai & Trenberth, 2002) or user-provided custom files.
 10. **CDR Forcing:** User-defined interventions that inject BGC tracers at point sources or as larger-scale Gaussian perturbations to simulate CDR interventions. The CDR forcing is prescribed as volume and tracer fluxes (e.g., alkalinity for ocean alkalinity enhancement, iron for iron fertilization, or other BGC constituents). Users can control the magnitude, spatial footprint, and temporal evolution, allowing flexible representation of CDR interventions.
- Some source datasets are accessed automatically by the package, including Natural Earth, Dai & Trenberth runoff, and ERA5 meteorology, while users must manually download SRTM15, GSHHG, GLORYS, the BGC datasets, and TPXO tidal files. Although these are the datasets currently supported, the package's modular design makes it straightforward to add new source datasets in the future.
- To generate the model inputs, ROMS-Tools automates several intermediate processing steps, including:
- **Bathymetry processing:** The bathymetry is smoothed in two stages, first across the entire model domain and then locally in areas with steep slopes, to ensure local steepness ratios do not exceed a prescribed threshold in order to reduce pressure-gradient errors. A minimum depth is enforced to prevent water levels from becoming negative during large tidal excursions.
 - **Mask definition:** The land-sea mask is generated by comparing the ROMS grid's horizontal coordinates with a coastline dataset using the regionmask package (Hauser et al., 2024). Enclosed basins are subsequently filled with land.
 - **Land value handling:** Land values are filled via an algebraic multigrid method using pyamg (Bell et al., 2023) prior to horizontal regridding. This extends ocean values into land areas to resolve discrepancies between source data and ROMS land masks, preventing land-originating values from appearing in ocean cells.
 - **Regridding:** Ocean and atmospheric fields are horizontally and vertically regridded from standard latitude-longitude-depth grids to the model's curvilinear grid with a terrain-following vertical coordinate using xarray (Hoyer & Hamman, 2017) and xgcm (Busecke & contributors, 2025). Velocities are rotated to align with the curvilinear ROMS grid.
 - **Longitude conventions:** ROMS-Tools handles differences in longitude conventions, converting between [-180°, 180°] and [0°, 360°] as needed.
 - **River locations:** Rivers that fall within the model domain are automatically identified and relocated to the nearest coastal grid cell. Rivers that need to be shifted manually or

span multiple cells can be configured by the user.

- **Data streaming:** ERA5 atmospheric data can be accessed directly from the cloud, removing the need for users to pre-download large datasets locally. Similar streaming capabilities may be implemented for other datasets in the future.

Users can quickly design and visualize regional grids and inspect all input fields with built-in plotting utilities. An example of surface initial conditions generated for a California Current System simulation at 5 km horizontal grid spacing is shown in [Figure 1](#).

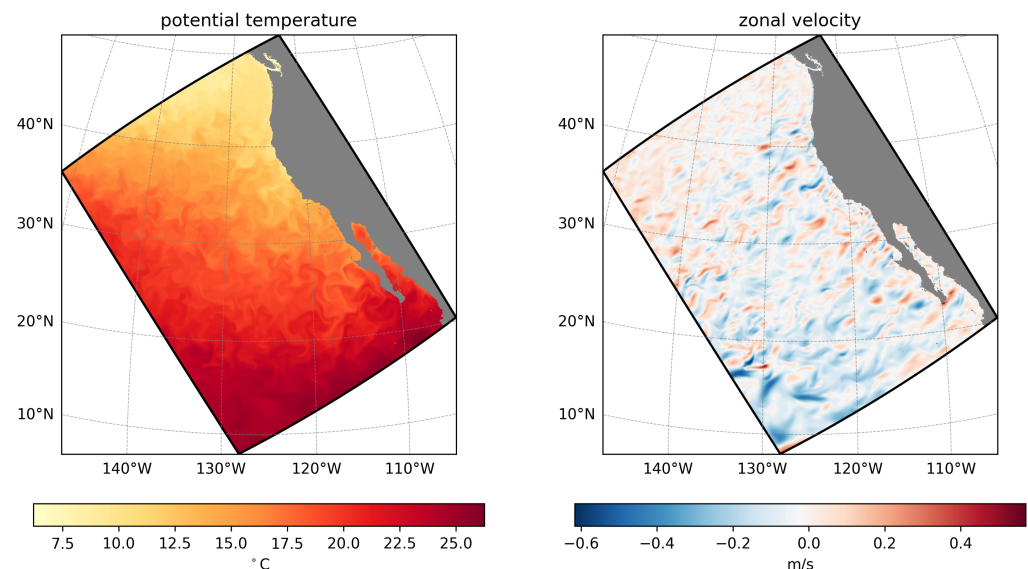


Figure 1: Surface initial conditions for the California Current System created with ROMS-Tools from GLORYS. Left: potential temperature. Right: zonal velocity. Shown for January 1, 2000.

Postprocessing and Analysis

ROMS-Tools supports postprocessing and analysis of ROMS-MARBL output, including regridding from the native curvilinear, terrain-following grid to a standard latitude-longitude-depth grid using *xesmf* ([Zhuang et al., 2023](#)), with built-in plotting for both grid types. The analysis layer also includes specialized utilities for evaluating carbon dioxide removal (CDR) interventions, such as generating carbon uptake and efficiency curves.

Statement of Need

Setting up a regional ocean model is a major technical undertaking. Traditionally, this work has relied on a patchwork of custom scripts and lab-specific workflows, which can be time-consuming, error-prone, and difficult to reproduce. These challenges slow down science, create a steep barrier to entry for new researchers, and limit collaboration across groups.

Within the ROMS community, the preprocessing landscape has been shaped by tools like *pyroms* ([Hedstrom & contributors, 2023](#)). While providing valuable low-level utilities, *pyroms* presents challenges for new users: installation is cumbersome due to Python/Fortran dependencies, the API is inconsistent, documentation is limited, and it lacks tests, CI, and support for modern Python tools like *xarray* and *dask*. Since active development has largely ceased, its suitability for new projects, such as CDR simulations, is limited.

Tools from other modeling communities cannot simply be adopted, since each ocean model has distinct structural requirements. For example, the *regional-mom6* package ([Barnes et al.,](#)

221 2024), developed for the Modular Ocean Model v6 (MOM6) (Adcroft et al., 2019), cannot be
222 used to generate ROMS inputs, because ROMS employs a terrain-following vertical coordinate
223 system that requires a specialized vertical regridding approach, whereas MOM6 accepts inputs
224 on arbitrary depth levels and does not require vertical regridding at all. Several other differences
225 further prevent cross-compatibility. Together, these limitations underscored the need for a
226 modern, maintainable, and reproducible tool designed specifically for ROMS.¹

227 ROMS-Tools was developed to meet this need. It draws on the legacy of the MATLAB
228 preprocessing scripts developed at UCLA (Molemaker, 2024), which encapsulate decades of
229 expertise in configuring regional ocean model inputs. While many of the core algorithms and
230 design principles are retained, ROMS-Tools provides an open-source Python implementation of
231 these MATLAB tools using an object-oriented programming paradigm. This implementation
232 supports a modern workflow with high-level API calls, improving reproducibility, minimizing
233 user errors, and allowing easy extension to new features, forcing datasets, and use cases. In
234 some cases, ROMS-Tools diverges from the MATLAB implementation to take advantage of
235 new methods or better integration with the modern Python ecosystem. By streamlining input
236 generation and analysis, ROMS-Tools reduces technical overhead, lowers the barrier to entry,
237 and enables scientists to focus on research rather than data preparation.

238 Software Design

239 ROMS-Tools is designed to balance **ease of use, flexibility, reproducibility, and scalability** by
240 combining high-level user interfaces with a modular, extensible architecture. A key goal is to
241 reduce barriers for new users, addressing recurring pain points observed in the legacy MATLAB
242 preprocessing scripts developed at UCLA.

243 Lessons from MATLAB Tools

244 The original MATLAB scripts were powerful but required users to edit source code directly
245 to configure simulations. This workflow led to frequent errors for new users, made it difficult
246 to track completed steps, and limited reproducibility. ROMS-Tools addresses these issues with
247 **high-level API calls**, automated error-prone steps, and explicit workflow state management via
248 YAML.

249 Design Trade-Offs

250 A central design trade-off in ROMS-Tools is between **automation** and **user control**. Rather than
251 enforcing a fixed workflow, the package exposes key choices such as physical options (e.g.,
252 radiation or wind corrections), interpolation and fill methods, and computational backends.
253 This approach contrasts with opinionated frameworks that fix defaults and directory structures
254 to maximize automation. While users must make explicit decisions, some steps remain
255 automated to prevent errors. For example, bathymetry smoothing is applied automatically with
256 a non-tunable parameter, since overly small smoothing factors could produce rough bathymetry
257 and crash simulations due to pressure gradient errors. This design choice directly addresses
258 issues new users faced in the MATLAB scripts, and balances **flexibility, reproducibility, and**
259 **safety**, enabling transparent experimentation without exposing users to avoidable pitfalls.

260 Another key trade-off is between **monolithic workflows** and **incremental, modular steps**. ROMS-
261 Tools uses small, composable components (such as generating initial conditions, boundary
262 forcing, and surface forcing) that can be executed, saved, and revisited independently. This
263 avoids unnecessary recomputation when only some inputs change. To ensure reproducibility
264 despite a modular workflow, configuration choices are stored in compact, text-based YAML
265 files. These files are version-controllable, easy to share, and remove the need to transfer large

¹In the future, packages like ROMS-Tools and regional-mom6 could share a common backbone, with model-specific adaptations layered on top.

166 model input NetCDF datasets. This directly remedies the MATLAB scripts' lack of explicit
167 workflow tracking.

168 Architecture

169 At the user-facing level, ROMS-Tools provides high-level objects such as `Grid`, `InitialConditions`,
170 and `BoundaryForcing`. Each object exposes a consistent interface (`.ds`, `.plot()`, `.save()`,
171 `.to_yaml()`), allowing users to call the same methods in sequence or inspect attributes that
172 are always present. This design reduces cognitive overhead, makes workflows predictable
173 and easy to follow, and eliminates the need for new users to manipulate raw scripts or track
174 intermediate files manually.

175 Internally, ROMS-Tools uses a **layered, modular architecture**. Abstract base classes
176 (`LatLonDataset`, `ROMSDataset`) handle data ingestion and preprocessing. Source-specific
177 datasets (e.g., `ERA5Dataset`, `GLORYSDataset`, `SRTMDataset`) inherit from these base classes
178 and encode dataset-specific conventions, such as variable names, coordinates, and masking.
179 Common operations, like subdomain selection and lateral filling, are implemented once and
180 reused across datasets. Adding a new data source typically requires only a small subclass to
181 define variable mappings while reusing the existing logic, keeping changes to the core code
182 minimal. This layered design supports **extensibility and maintainability**, avoiding the pitfalls of
183 the monolithic MATLAB scripts.

184 Computational and Data Model Choices

185 ROMS-Tools is built on `xarray`, which provides a clear, consistent interface for exploring and
186 inspecting labeled, multi-dimensional geophysical datasets. Users can take advantage of
187 `xarray`'s intuitive indexing, plotting, and metadata handling. Optional dask support allows
188 workflows to scale from laptops to HPC systems, enabling parallel and out-of-core computation
189 for very large input and output datasets. By combining modern Python tools with a user-friendly
190 interface, ROMS-Tools addresses the usability and reproducibility challenges that hampered
191 new users in the MATLAB-based workflow.

192 Research Impact Statement

193 ROMS-Tools is used by two primary research communities. First, regional ocean modelers
194 use it to generate reproducible input datasets for ROMS simulations; external users include
195 researchers at **PNNL**, **WHOI**, and **UCLA**. Second, researchers in the ocean-based carbon
196 dioxide removal (CDR) community use ROMS-Tools to configure reproducible ROMS-MARBL
197 simulations of climate intervention scenarios, with adopters including **[C]Worthy**, **Carbon to**
198 **Sea**, **Ebb Carbon**, and **SCCWRP**. All of these groups have contacted the developers directly
199 or engaged in offline discussions regarding their use of the package.

200 Additional evidence of community uptake comes from public usage metrics. At the time of
201 writing, the GitHub repository shows **119 unique cloners in the past 14 days**, with stars
202 from users at institutions including the University of Waikato, NCAR, University of Maryland,
203 National Oceanography Centre, McGill University, UC Santa Cruz, and others. Distribution
204 statistics indicate **over 3,100 conda-forge downloads in the past six months**, including **68**
205 **downloads of the most recent release (v3.3.0)**, and **more than 48,000 total PyPI downloads**
206 (noting that PyPI counts include automated CI usage, whereas conda downloads do not).

207 ROMS-Tools is also integrated into broader workflows, including **C-Star**([Stephenson &](#)
208 [contributors, 2025](#)), an open-source platform to provide scientifically credible monitoring,
209 reporting, and verification (MRV) for the emerging marine carbon market.

AI Usage Disclosure

Generative AI tools were used to assist with writing docstrings and developing tests for the ROMS-Tools software, to improve the clarity and readability of the documentation, and to shorten and edit portions of the manuscript text. All AI-assisted content was reviewed and verified by the authors for technical accuracy and correctness.

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