

SpeedyWeather.jl: Reinventing atmospheric general circulation models towards interactivity and extensibility

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Summary

SpeedyWeather.jl is a library to simulate and analyze the global atmospheric circulation on the sphere. It implements several 2D and 3D models which solve different sets of equations:

- the primitive equations with and without humidity ([Figure 1](#)),
- the shallow water equations ([Figure 2](#)), and
- the barotropic vorticity equation.

The primitive equation model in SpeedyWeather.jl is an atmospheric general circulation model ([Kucharski et al., 2013](#)) with simple parameterizations for unresolved physical processes including precipitation or boundary layer mixing. It can be thought of as a conceptual reinvention of the Fortran SPEEDY model ([Molteni, 2003](#)) in the Julia programming language ([Bezanson et al., 2017](#)). However, all models here are written in a modular way to make its components easily extensible. For example, a new parameterization can be externally defined and passed as an argument to the model constructor. Operators used inside SpeedyWeather.jl are exposed to the user, facilitating analysis of the simulation data. SpeedyWeather.jl is therefore, beyond its main purpose of simulating atmospheric motion, also a library for the analysis of gridded data on the sphere. Running and analyzing simulations can be interactively combined, enhancing user experience and productivity.

The user interface of SpeedyWeather.jl is heavily influenced by the Julia ocean model Oceananigans.jl ([Ramadhan et al., 2020](#)). A monolithic interface ([Mazlami et al., 2017](#)), controlling most of the model's functionality through arguments of a single function or through parameter files (often called namelists in Fortran), is avoided in favor of a library-style interface. A model is constructed bottom-up by first defining the discretization and any non-default model components with their respective parameters. All components are then collected into a single model object which, once initialized, returns a simulation object. A simulation contains everything, the model with all parameters as constructed before but also all prognostic and diagnostic variables. Such a simulation can then be run, but also accessed before and after to analyze or visualize the current variables, or individual terms of the equations. One can also adjust some parameters before resuming the simulation. While these steps can be written into a script for reproducibility, the same steps can be executed and interacted with one-by-one in Julia's read-evaluate-print loop (REPL) or in a Jupyter or Pluto notebook. We thereby achieve an

43 interactivity of a simulation and its various model components far beyond the options provided
44 in a monolithic interface. At the same time, defaults, set to well-established test cases, enable
45 even inexperienced users to run simulations in just a few lines of code.

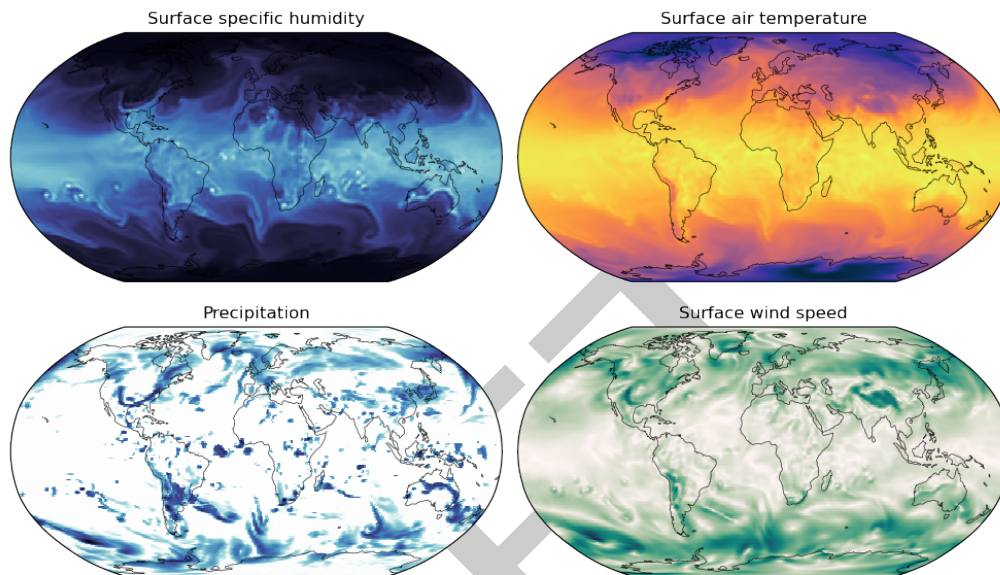
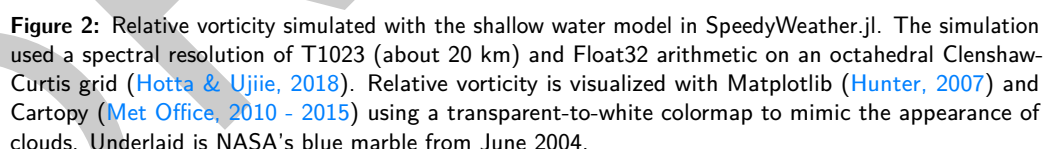


Figure 1: Surface humidity, air temperature, wind speed and precipitation simulated with the primitive equation model in SpeedyWeather.jl. Spectral resolution is T127 (about 100km) on an octahedral Gaussian grid (Malardel et al., 2016) with simple physics to represent unresolved processes such as surface fluxes including evaporation, and precipitation due to large-scale condensation and convection.

46 SpeedyWeather.jl relies on Julia's multiple dispatch programming paradigm (Bezanson et al.,
47 2017) to be extensible with new components including parameterizations, forcing, drag, or even
48 the grid. All such supported model components define an abstract type that can be subtyped to
49 introduce, for example, a new parameterization. To define a new parameterization for convection
50 in a given vertical column of the atmosphere, one would define MyConvection as a new subtype
51 of AbstractConvection. One then only needs to extend the initialize! (executed once
52 during model initialization) and convection! (executed on every time step) functions for
53 this new type. Passing on convection = MyConvection() to the model constructor then
54 implements this new model component without the need to branch off or overwrite existing
55 model components. Conceptually similar scientific modelling paradigms have been very
56 successful in the Python-based generic partial differential equation solver Dedalus (Burns et
57 al., 2020), the process-oriented climate model CLIMLAB (Rose, 2018), and the Julia ocean
58 model Oceananigans.jl (Ramadhan et al., 2020).

59 The dynamical core of SpeedyWeather.jl uses established numerics (Bourke, 1972; Hoskins
60 & Simmons, 1975; Simmons et al., 1978; Simmons & Burridge, 1981), widely adopted in
61 numerical weather prediction. It is based on the spherical harmonic transform (Reinecke &
62 Seljebo, 2013; Stompor, 2011) with a leapfrog-based semi-implicit time integration (Hoskins
63 & Simmons, 1975) and a Robert-Asselin-Williams filter (Amezcu et al., 2011; Williams, 2011).
64 The spherical harmonic transform is grid-flexible (Willmert, 2020). Any iso-latitude ring-based
65 grid can be used and new grids can be externally defined and passed in as an argument. Many
66 grids are already implemented: the conventional Gaussian grid, a regular longitude-latitude
67 grid, the octahedral Gaussian grid (Malardel et al., 2016), the octahedral Clenshaw-Curtis grid
68 (Hotta & Ujiie, 2018), and the HEALPix grid (Górski et al., 2005). Both SpeedyWeather.jl and
69 its spherical harmonic transform are also number format-flexible. Single-precision floating-point
70 numbers (Float32) are the default as adopted by other modelling efforts (Nakano et al., 2018;

SpeedyWeather.jl internally uses three sub-modules RingGrids, LowerTriangularMatrices, and SpeedyTransforms. RingGrids is a module that discretizes the sphere on iso-latitude rings and implements interpolations between various such grids. LowerTriangularMatrices facilitates the implementation of the spherical harmonics by organizing their coefficients in a lower triangular matrix representation. SpeedyTransforms implements the spectral transform between the grid-point space as defined by RingGrids and the spectral space defined in LowerTriangularMatrices. These three modules are independently usable and therefore support SpeedyWeather's library-like user interface. Output is stored as NetCDF files using NCDatasets.jl(Barth, 2023).



SpeedyWeather.jl is a fresh approach to atmospheric models that have been very influential in many areas of scientific and high-performance computing as well as climate change mitigation and adaptation. Most weather, ocean and climate models are written in Fortran (e.g. ICON (Giorgetta et al., 2018), CESM (Hurrell et al., 2013), MITgcm (Marshall et al., 1997), NEMO (Madec et al., 2017)) and have been developed over decades. From this tradition follows a specific programming style and associated user interface. SpeedyWeather.jl aims to overcome the constraints of traditional Fortran-based models. The modern trend sees simulations in Fortran and data analysis in Python (e.g. NumPy (Harris et al., 2020), Xarray (Hoyer & Hamman, 2017), Dask (Dask Development Team, 2016), Matplotlib (Hunter, 2007)), making it virtually impossible to interact with various model components directly. In SpeedyWeather.jl, interfaces to the model components are exposed to the user. Furthermore, data-driven climate modelling (Rasp et al., 2018; Schneider et al., 2023), which replaces existing model components with machine learning, is more difficult in Fortran due to the lack of established machine

learning frameworks (Meyer et al., 2022). In Julia, Flux.jl (Innes et al., 2019) is available for machine learning as well as automatic differentiation with Enzyme (Moses & Churavy, 2020) for gradients-based optimization.

With SpeedyWeather.jl we hope to provide a platform for data-driven atmospheric modelling and in general an interactive model that makes difficult problems easy to simulate. Climate models that are user-friendly, trainable, but also easily extensible will suddenly make many complex research ideas possible.

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