

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

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Summary

SpeedyWeather.jl is a library to simulate and analyse the global atmospheric circulation on the sphere. It implements several 2D and 3D models to solve the primitive equations with and without humidity (Figure 1), the shallow water equations (Figure 2), or the barotropic vorticity equations with spherical harmonics. Several simple parameterizations for unresolved physical processes such as precipitation or the boundary layer are implemented, and new ones can be externally defined and passed as an argument to the model constructor. SpeedyWeather.jl is an intermediate-complexity general circulation model (Kucharski et al., 2013) and research playground with an (almost) everything-flexible attitude. 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).

SpeedyWeather.jl internally uses three sub-modules SpeedyTransforms, RingGrids, and LowerTriangularMatrices to perform spherical harmonic transforms and interpolations between various grids and the spectral space. RingGrids discretize the sphere on iso-latitude rings and the spectral space is defined by the LowerTriangularMatrices of the spherical harmonic coefficients. These three modules are independently usable and therefore make SpeedyWeather.jl, beyond its main purpose of simulating the weather, also a library for the analysis of gridded data on the sphere. Running and analysing 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 based on parameter files is avoided in favor of a library-style interface in which users write code to run models rather than merely supplying parameters and input arrays. A model is created bottom-up by first defining the discretization and any non-default model components with its respective parameters. All components are then collected into a single model object, which, once initialized, returns a simulation object that contains the entire model state, work arrays and parameters, that can be run, analysed or changed. 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). We thereby reach an interactivity far beyond a monolithic interface that is limited to the options provided.

43 To be extensible and composable with new model components, SpeedyWeather.jl relies on Julia's
44 multiple dispatch programming paradigm (Bezanson et al., 2017). Every model component is
45 defined as a new type. For example, to define a new way how to calculate the precipitation
46 due to the physical process of large-scale condensation, one would define MyCondensation as a
47 new subtype of AbstractCondensation. One then only needs to extend the initialize! and
48 condensation! functions for this new type. Passing on condensation = MyCondensation()
49 to the model constructor then implements this new model component without the need to
50 branch off or overwrite existing model components. Conceptually similar scientific modelling
51 paradigms have been very successful in the Python-based generic partial differential equation
52 solver Dedalus (Burns et al., 2020) and the Julia ocean model Oceananigans.jl (Ramadhan et
53 al., 2020).

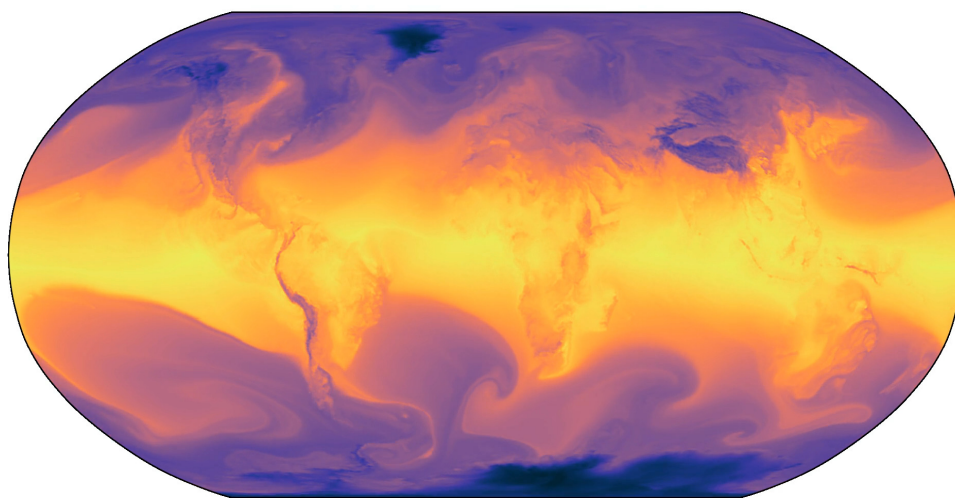


Figure 1: Surface temperature simulated with the primitive equation model in SpeedyWeather.jl. (Figure will be updated)

54 The dynamical core of SpeedyWeather.jl uses established numerics (Bourke, 1972; Hoskins
55 & Simmons, 1975; Simmons et al., 1978; Simmons & Burridge, 1981), widely adapted in
56 numerical weather prediction. It is based on the spherical harmonic transform with a leapfrog-
57 based semi-implicit time integration (Hoskins & Simmons, 1975) and a Robert-Asselin-Williams
58 filter (Amezcuca et al., 2011; Williams, 2011). The spherical harmonic transform is grid-flexible.
59 Any iso-latitude ring-based grid can be used and new grids can be externally defined and
60 passed on as argument. Many grids are already implemented: the conventional Gaussian grid,
61 a regular longitude-latitude grid, the octahedral Gaussian grid (Malardel et al., 2016), the
62 octahedral Clenshaw-Curtis grid (Hotta & Ujiie, 2018), and the HEALPix grid (Górski et al.,
63 2005). Both SpeedyWeather.jl and its spherical harmonic transform SpeedyTransforms are
64 also number format-flexible. 32-bit single-precision floating-point numbers (Float32) are the
65 default as adapted by other modelling efforts Nakano et al. (2018), but Float64 and other
66 custom number formats can be used with a single code basis. Julia will compile to the choice
67 of the number format, the grid, and and other model components just-in-time. A simple
68 parallelisation across vertical layers is supported with Julia's multi threading.

69 Statement of need

70 SpeedyWeather.jl is a fresh approach to atmospheric models that have been very influential in
71 many areas of scientific and high-performance computing as well as climate change mitigation
72 and adaptation. Most weather, ocean and climate models are written in Fortran and have been
73 developed over decades. From this tradition follows a specific programming style and associated

74 user interface. Running a simulation in Fortran and analysing the data in Python makes it
75 virtually impossible to interact with various model components interactively. Furthermore,
76 data-driven climate modelling (Rasp et al., 2018, p. Schneider2023), which replaces existing
77 model components with machine learning is difficult due to the lack of established deep learning
78 frameworks in Fortran (Innes et al., 2019). Let alone online learning, which trains a neural
79 network-based component together with the rest of the model, accounting for interactions
80 between components. Gradients, necessary to optimize training, can be computed with
81 automatic differentiation (Moses & Churavy, 2020), but only if differentiable functions in a
82 coherent language framework are provided.

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

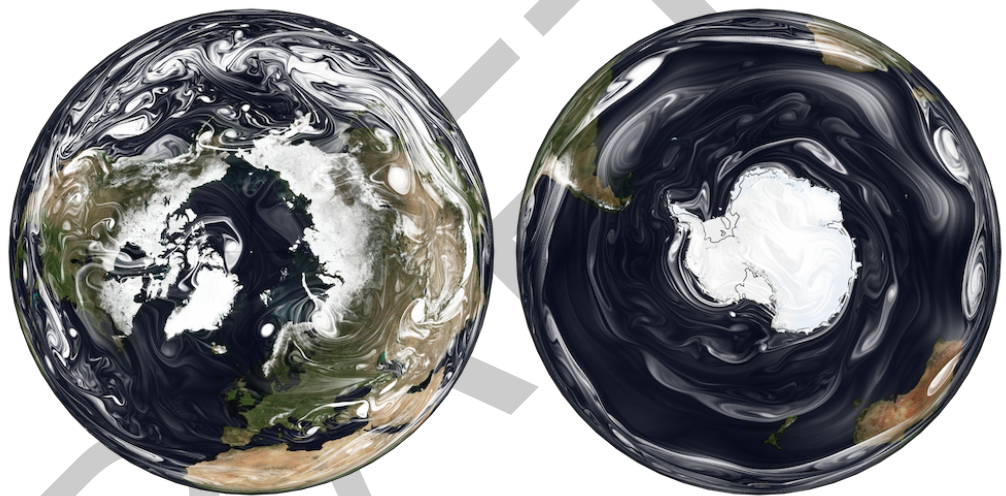


Figure 2: Relative vorticity simulated with the shallow water model in SpeedyWeather.jl. The simulation used a spectral resolution of T1023 (about 20km) and Float32 arithmetic on an octahedral Clenshaw-Curtis grid (Hotta & Ujiie, 2018). Relative vorticity is visualised with Matplotlib (Hunter, 2007) and Cartopy (Met Office, 2010 - 2015) using a transparent-to-white colormap to mimic the appearance of clouds. Underlain is NASA's blue marble from June 2004.

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