

- PyMPDATA v1: Numba-accelerated implementation of MPDATA with examples in Python, Julia and Matlab
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Software

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Statement of need

Convection-diffusion problems arise across a wide range of pure and applied research, in particular in geosciences, aerospace engineering, and financial modelling (for an overview of applications, see, e.g., section 1.1 in Morton (1996)). One of the key challenges in numerical solutions of problems involving advective transport is sign preservation of the advected field (for an overview of this and other aspects of numerical solutions to advection problems, see, e.g., Røed (2019)). The Multidimensional Positive Definite Advection Transport Algorithm (MPDATA) is a robust, explicit-in-time, and sign-preserving solver introduced in Smolarkiewicz (1983) and Smolarkiewicz (1984). MPDATA has been subsequently developed into a family of numerical schemes with numerous variants and solution procedures addressing a diverse set of problems in geophysical fluid dynamics and beyond. For reviews of MPDATA applications and variants, see, e.g., Smolarkiewicz & Margolin (1998) and Smolarkiewicz (2006).

The PyMPDATA project introduced herein constitutes a high-performance and multi-threaded implementation of structured-mesh MPDATA in Python. PyMPDATA is aimed to address several aspects which steepen the learning curve and limit collaborative usage and development of existing C++ (e.g., Jaruga et al., 2015) and Fortran (e.g., Kühnlein et al., 2019) implementations of MPDATA. Performance on par with compiled-language implementations is targeted by employing just-in-time (JIT) compilation using Numba (Lam et al., 2015), which translates Python code into fast machine code using the Low Level Virtual Machine (LLVM, https://llvm.org/) compiler infrastructure (for a discussion of another JIT implementation of MPDATA using PyPy, see Arabas et al., 2014). PyMPDATA is engineered aiming at both performance and usability, the latter encompassing research user's, developer's and maintainer's perspectives. From researcher's perspective, PyMPDATA offers hassle-free installation on Linux, macOS, and Windows. It also eliminates the compilation stage from the perspective of the user. From developer's and maintainer's perspectives, PyMPDATA offers a suite of unit tests, multi-platform continuous integration setup, seamless integration with Python development tools including debuggers, profilers, and code analysers.

Summary

PyMPDATA interface uses NumPy for array-oriented input and output. Usage of PyMPDATA from
Julia (https://julialang.org) and Matlab (https://mathworks.com) through PyCall and the
built-in Python interoperability tools, respectively, is depicted in the PyMPDATA README
file. The appendices of the present paper include Python, Julia and Matlab minimal code
snippets covering steps needed to complete a basic PyMPDATA simulation depicted in Fig.
(Figure 1) and based on Fig. 5 from Arabas et al. (2014).



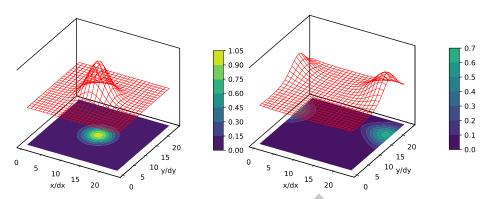


Figure 1: Visualisation of the initial condition (left) and simulation state after 75 timesteps (right) from the 2D simulation for which sample codes are given in the appendices, based on Fig. 5 from Arabas et al. (2014).

As of the current version, PyMPDATA supports homogeneous transport in one (1D), two (2D), and three dimensions (3D) using structured meshes, optionally generalised by coordinate transformation (Smolarkiewicz & Clark, 1986; Smolarkiewicz & Margolin, 1993). PyMPDATA includes implementation of a subset of MPDATA variants, such as the non-oscillatory option (Smolarkiewicz & Grabowski, 1990), the infinite-gauge variant (Margolin & Shashkov, 2006; Smolarkiewicz & Clark, 1986), the divergent-flow option (Margolin & Smolarkiewicz, 1998; Smolarkiewicz, 1984), the double-pass donor cell (DPDC) flavour (Beason & Margolin, 1988; Margolin & Shashkov, 2006; Margolin & Smolarkiewicz, 1998), and the third-order-terms options (Margolin & Smolarkiewicz, 1998). It also features support for integration of Fickianterms in advection-diffusion problems using the pseudo-transport velocity approach (Smolarkiewicz & Clark, 1986; Smolarkiewicz & Szmelter, 2005).

A companion package named PyMPDATA-examples contains a set of Jupyter notebooks, which reproduce results from literature using PyMPDATA. These examples are also executed within continuous integration runs. Several of the examples feature comparisons against analytical solution, and these are also included in the test suite of PyMPDATA. The PyMPDATA-examples README file includes links (badges) offering single-click deployment in the cloud using either the Binder (https://mybinder.org) or the Colab (https://colab.research.google.c om) platforms.

A separate project named numba-mpi has been developed to set the stage for future Message Passing Interface (MPI) distributed memory parallelism in PyMPDATA. The PyMPDATA, the PyMPDATA-examples and the numba-mpi packages are available in the PyPI package repository, and installation of these packages reduces to typing pip install package_name. Development of all three packages is hosted on GitHub at: https://github.com/atmos-cloud-sim-uj/ and continuous integration runs on Linux, macOS and Windows using GitHub Actions and Appveyor platforms (the latter used for 32-bit runs on Windows). Auto-generated documentation sites built with pdoc3 are hosted at https://atmos-cloud-sim-uj.github.io/PyMPDATA/, https://atmos-cloud-sim-uj.github.g

are hosted at https://atmos-cloud-sim-uj.github.io/PyMPDATA/, https://atmos-cloud-sim-uj.github.io/PyMPDATA-examples/, and https://atmos-cloud-sim-uj.github.io/numba-mpi/.

PyMPDATA is a free and open-source software released under the terms of the GNU General Public License 3.0 (https://www.gnu.org/licenses/gpl-3.0).

Usage examples

Simulations included in the PyMPDATA-examples package are listed below, labelled with the paper reference on which the example setup is based. Each example is annotated with the dimensionality, number of equations constituting the system, and an outline of setup.



■ 1D:

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- Smolarkiewicz (2006): single-equation advection-only homogeneous problem with different algorithm options depicted with constant velocity field
- Arabas & Farhat (2020): single-equation advection-diffusion problem resulting from a transformation of the Black-Scholes equation into either homogeneous or heterogeneous problem for European or American option valuation, respectively
- Olesik et al. (2021): single-equation advection-only homogeneous problem with coordinate transformation depicting application of MPDATA for condensational growth of a population of particles

■ 2D:

- Molenkamp (1968): single-equation homogeneous transport with different algorithm options
- Jarecka et al. (2015): shallow-water system with three equations representing conservation of mass and two components of momentum (with the momentum equations featuring source terms) modelling spreading under gravity of a threedimensional elliptic drop on a two-dimensional plane
- Williamson & Rasch (1989): advection on a spherical plane depicting transformation to spherical coordinates
- Shipway & Hill (2012): coupled system of water vapour mass (single spatial dimension) and water droplet number conservation (spatial and spectral dimensions) with the latter featuring source term modelling activation of water droplets on aerosol particles, coordinate transformation used for representation of air density profile

■ 3D:

 Smolarkiewicz (1984): homogeneous single-equation example depicting revolution of a spherical signal in a constant angular velocity rotational velocity field

In addition, PyMPDATA is used in a two-dimensional setup in of the examples in the sister PySDM package (Bartman et al., 2021).

Implementation highlights

In 2D and 3D simulations, domain-decomposition is used for multi-threaded parallelism. Domain decomposition is performed along the outer dimension only and is realised using the numba.prange() functionality.

PyMPDATA design features a custom-built multi-dimensional Arakawa-C staggered grid layer, allowing to concisely represent multi-dimensional stencil operations on both scalar and vector fields. The grid layer is built on top of NumPy's ndarrays (using "C" ordering) using the Numba's @njit functionality for high-performance multi-threaded array traversals. The array-traversal layer enables to code once for multiple dimensions (i.e. one set of MPDATA formulae for 1D, 2D, and 3D), and automatically handles (if needed) any halo-filling logic related to boundary conditions.

The Numba's deviation from Python semantics rendering closure variables as compile-time constants is extensively exploited within PyMPDATA code base enabling the just-in-time compilation to benefit from information on domain extents, algorithm variant used and problem characteristics (e.g., coordinate transformation used, or lack thereof).

In general, the numerical and concurrency aspects of PyMPDATA implementation follow the libmpdata++ open-source C++ implementation of MPDATA (Jaruga et al., 2015).



119 Appendix P: Python sample code

```
import numpy as np
from PyMPDATA import Options, ScalarField, VectorField, Stepper, Solver
from PyMPDATA.boundary conditions import Periodic
options = Options(n_iters=2)
nx, ny = 24, 24
Cx, Cy = -.5, -.25
halo = options.n_halo
xi, yi = np.indices((nx, ny), dtype=float)
advectee = ScalarField(
  data=np.exp(
    -(xi+.5-nx/2)**2 / (2*(nx/10)**2)
    -(yi+.5-ny/2)**2 / (2*(ny/10)**2)
  ),
  halo=halo,
  boundary_conditions=(Periodic(), Periodic())
advector = VectorField(
  data=(np.full((nx + 1, ny), Cx), np.full((nx, ny + 1), Cy)),
  halo=halo,
  boundary_conditions=(Periodic(), Periodic())
stepper = Stepper(options=options, grid=(nx, ny))
solver = Solver(stepper-stepper, advectee=advectee, advector=advector)
solver.advance(n_steps=75)
state = solver.advectee.get()
```

120 Appendix J: Julia sample code

```
using PyCall
Options = pyimport("PyMPDATA").Options
ScalarField = pyimport("PyMPDATA").ScalarField
VectorField = pyimport("PyMPDATA").VectorField
Stepper = pyimport("PyMPDATA").Stepper
Solver = pyimport("PyMPDATA").Solver
Periodic = pyimport("PyMPDATA.boundary_conditions").Periodic
options = Options(n_iters=2)
nx, ny = 24, 24
Cx, Cy = -.5, -.25
idx = CartesianIndices((nx, ny))
halo = options.n_halo
advectee = ScalarField(
    data=exp.(
        -(getindex.(idx, 1) .- .5 .- nx/2).^2 / (2*(nx/10)^2)
        -(getindex.(idx, 2) .- .5 .- ny/2).^2 / (2*(ny/10)^2)
    ),
    halo=halo,
    boundary_conditions=(Periodic(), Periodic())
)
```



```
advector = VectorField(
    data=(fill(Cx, (nx+1, ny)), fill(Cy, (nx, ny+1))),
    halo=halo,
    boundary_conditions=(Periodic(), Periodic())
)

stepper = Stepper(options=options, grid=(nx, ny))
solver = Solver(stepper=stepper, advectee=advectee, advector=advector)
solver.advance(n_steps=75)
state = solver.advectee.get()
```

21 Appendix M: Matlab sample code

```
Options = py.importlib.import_module('PyMPDATA').Options;
ScalarField = py.importlib.import_module('PyMPDATA').ScalarField;
VectorField = py.importlib.import module('PyMPDATA').VectorField;
Stepper = py.importlib.import_module('PyMPDATA').Stepper;
Solver = py.importlib.import_module('PyMPDATA').Solver;
Periodic = ...
    py.importlib.import_module('PyMPDATA.boundary_conditions').Periodic;
options = Options(pyargs('n_iters', 2));
nx = int32(24);
ny = int32(24);
Cx = -.5;
Cy = -.25;
[xi, yi] = meshgrid(double(0:1:nx-1), double(0:1:ny-1));
halo = options.n_halo;
advectee = ScalarField(pyargs(...
    'data', py.numpy.array(exp( ...
        -(xi+.5-double(nx)/2).^2 / (2*(double(nx)/10)^2) ...
        -(yi+.5-double(ny)/2).^2 / (2*(double(ny)/10)^2) ...
    )), ...
    'halo', halo, ...
    'boundary_conditions', py.tuple({Periodic(), Periodic()}) ...
advector = VectorField(pyargs(...
    'data', py.tuple({ ...
        Cx * py.numpy.ones(int32([nx+1 ny])), ...
        Cy * py.numpy.ones(int32([nx ny+1])) ...
     }), ...
    'halo', halo, ...
    'boundary_conditions', py.tuple({Periodic(), Periodic()}) ...
));
stepper = Stepper(pyargs(...
  'options', options, ...
  'grid', py.tuple({nx, ny}) ...
));
solver = Solver(...
```



```
pyargs('stepper', ...
    stepper, 'advectee', ...
    advectee, 'advector', ...
    advector ...
));
solver.advance(pyargs('n_steps', 75));
state = solver.advectee.get();
```

Author contributions

PB had been the architect of PyMPDATA with SA taking the role of main developer and 123 maintainer over the time. MO participated in the package core development and led the 124 development of the condensational-growth example, which was the basis of his MSc thesis. 125 JB contributed the DPDC algorithm variant handling. SD contributed the advection-diffusion example. MM contributed to the numba-mpi package. PR contributed the shallow-water 127 example. MS contributed the advection-on-a-sphere example. The paper was composed by 128 SA and is based on the contents of the README files of the PyMPDATA, PyMPDATA-examples, 129 and numba-mpi packages.

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