



Seeking portability and productivity for NWP model code

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Keynote: Emerging Technologies & Opportunities: GPU and Earth System Modeling



CSCS

Centro Svizzero di Calcolo Scientifico
Swiss National Supercomputing Centre



Platform for Advanced Scientific Computing

ETH zürich



UIFCW 2023

A UFS Collaboration Powered by **EPIC**



CENTRE OF EXCELLENCE IN SIMULATION OF WEATHER AND CLIMATE IN EUROPE



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Seeking portability and productivity

- ❖ Emerging technologies offer great potential for higher numerical resolution and energy efficiency. At the same time, ESMs face an increasingly diverse landscape of supercomputing architectures.
- ❖ Efficient execution requires targeted hardware-specific optimization. Serving various hardware inevitably involves more complex code that needs to be organized to maintain productivity.

Two parallel streams of development at ECMWF:

Operational IFS: Main ECMWF scalability & portability efforts prepare the spectral-transform forecast model for hybrid CPU+GPU execution. Automatic code translation tools are developed and employed, accompanied by restructuring core model components and various technical infrastructure packages. Fortran is largely maintained and GPU execution is enabled mostly by means of OpenACC directives.

Future IFS with new dynamical core: We are rewriting (from existing Fortran) and further [developing the forecast model in Python with the domain-specific library GT4Py](#), in close collaboration with partners at CSCS and ETH Zurich. The forecast model is building on finite-volume non-hydrostatic dynamical core with the IFS physical parametrizations.

GT4Py domain-specific library

- ❑ Software programming implementation using the domain-specific library GT4Py (Gridtools for Python) for stencil computations in grid-point NWP and climate models (Afanasyev et al. 2021).
- ❑ <https://github.com/GridTools/gt4py> (public, open source)
- ❑ GT4Py works as an optimizing compiler for multiple backends:
 - Code generation optimized for a specific architecture
 - Backend selects HPC implementation strategy (e.g., parallelization, memory layout, etc)
 - Backends can be added to provide efficient implementations for new technologies / architectures
 - **Leverages knowledge of the typical computation patterns in the domain**
- ❑ GT4Py is embedded in the **Python** eco-system
 - Portable and productive programming environment
 - Broad and comprehensive selection of modules and libraries
 - Most popular data scientist language
 - Low barrier of entry for domain scientists and academia

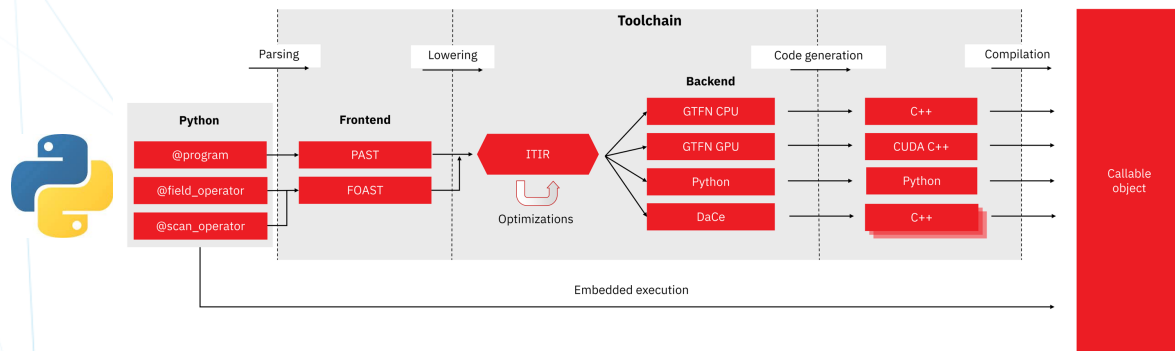
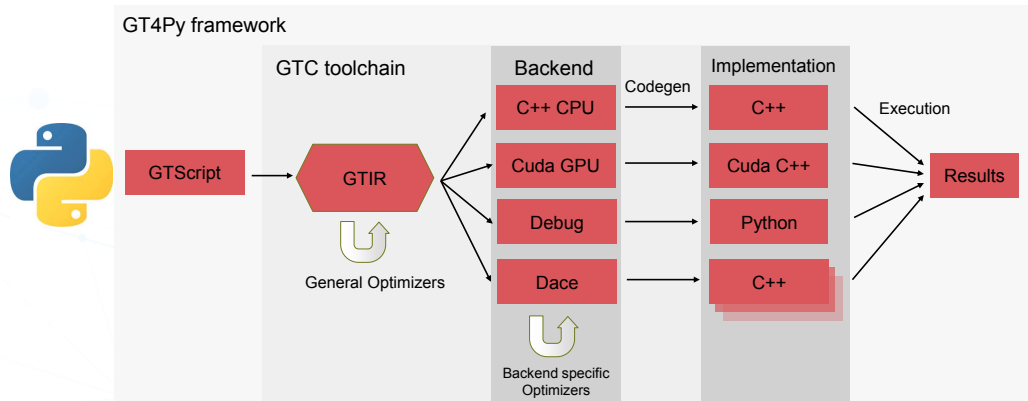


GT4Py domain-specific library

Established **GT4Py “Version 1”** for structured (I, J, K) grids.
See Afanasyev et al. 2021; Ben-Hun et al. 2022.

→ Results in subsequent slides are based on structured (I, J, K) grid model.

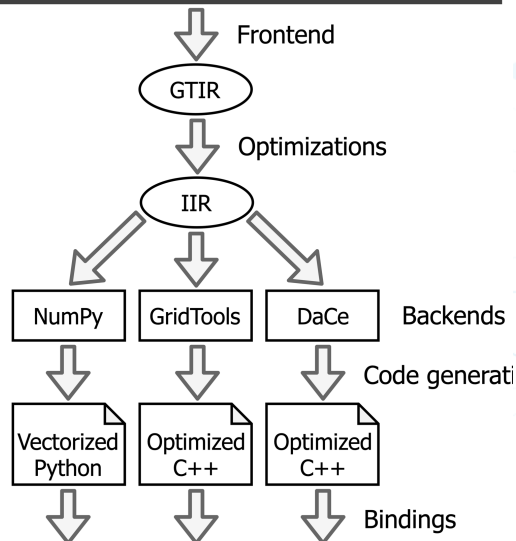
Declarative GT4Py with new interface & toolchain and supporting horizontally unstructured, e.g., (IJ, K), grids.
This is an ongoing development, see e.g. Bianco et al. PASC2023 poster.



GT4Py domain-specific library



```
@gtscript.stencil(backend="...")
def laplacian(
    in_phi: gtscript.Field[float], out_lap: gtscript.Field[float]
):
    with computation(PARALLEL), interval(...):
        out_lap[0, 0, 0] = - 4 * in_phi[0, 0, 0] \
            + in_phi[-1, 0, 0] + in_phi[1, 0, 0] \
            + in_phi[0, -1, 0] + in_phi[0, 1, 0]
```



```
laplacian(phi, lap, origin=(1, 1, 0), domain=(nx-2, ny-2, nz))
```

Schematic from
Ubbiali et al. in
prep. 2023

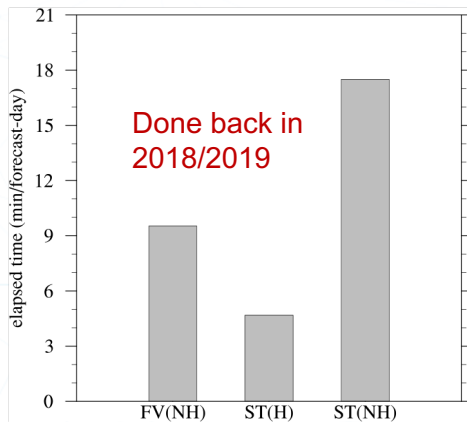
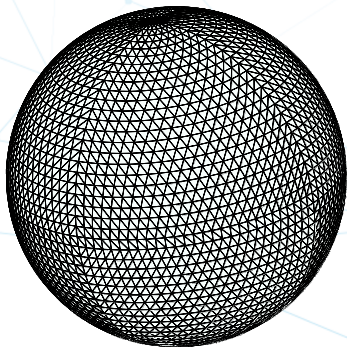
GT4Py domain-specific library

- ❑ Three comprehensive GT4Py based NWP and climate model software development projects (all coming from original Fortran implementations that were optimized for CPUs):
 - **Pace** is a GT4Py Version 1 implementation of the **FV3GFS / SHIELD** atmospheric model of NOAA and GFDL by Allen Institute for AI (AI2), ETH Zurich, and CSCS.
 - **ICON** atmospheric model dynamics and physics incrementally ported to declarative GT4Py by MeteoSwiss, EXCLAIM project at ETH Zurich and CSCS.
 - **IFS-FVM** porting and further development in Version 1 & declarative GT4Py by ECMWF, CSCS, and PASC-funded project KILOS at ETH Zurich.

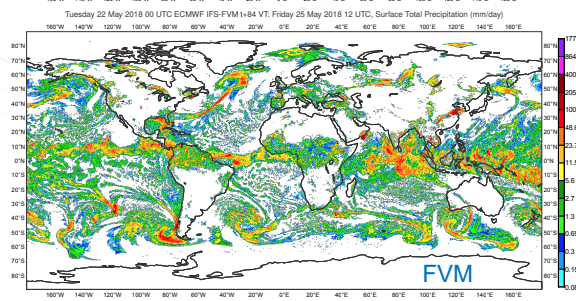
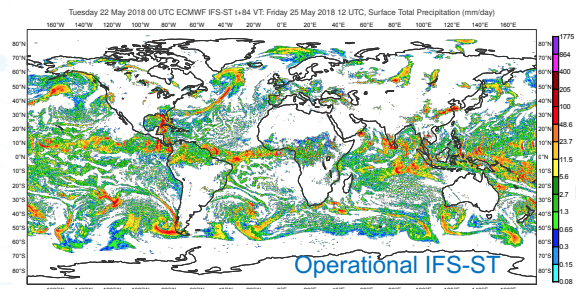
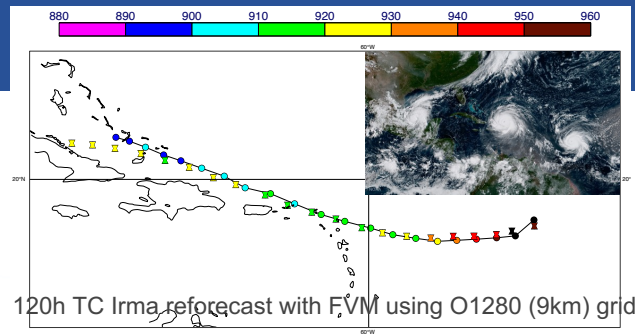
Starting point

- FVM original Fortran code with hybrid MPI & OpenMP parallelization targeting CPU based supercomputers
- Carefully optimized with good computational performance, e.g., against H and NH spectral-transform IFS for dynamical cores
- Operated on IFS octahedral grid provided by ECMWF Atlas library support

O24

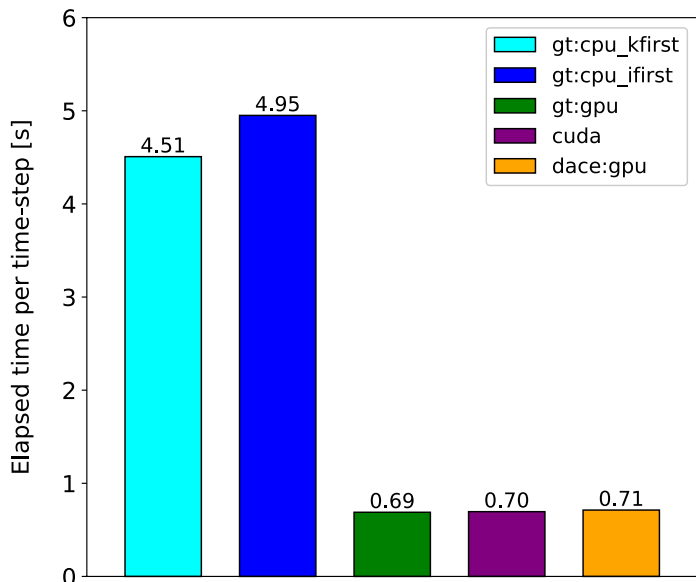


Dynamical core time-to-solution for DCMIP2016 baroclinic wave with TCo1279 / O1280 (9km) grid and L137 on 350 nodes (12800 cores) of ECMWF's former Cray XC40



Snapshot total precip at 84h

Seamless execution of structured-grid FVM with various GT4Py backends

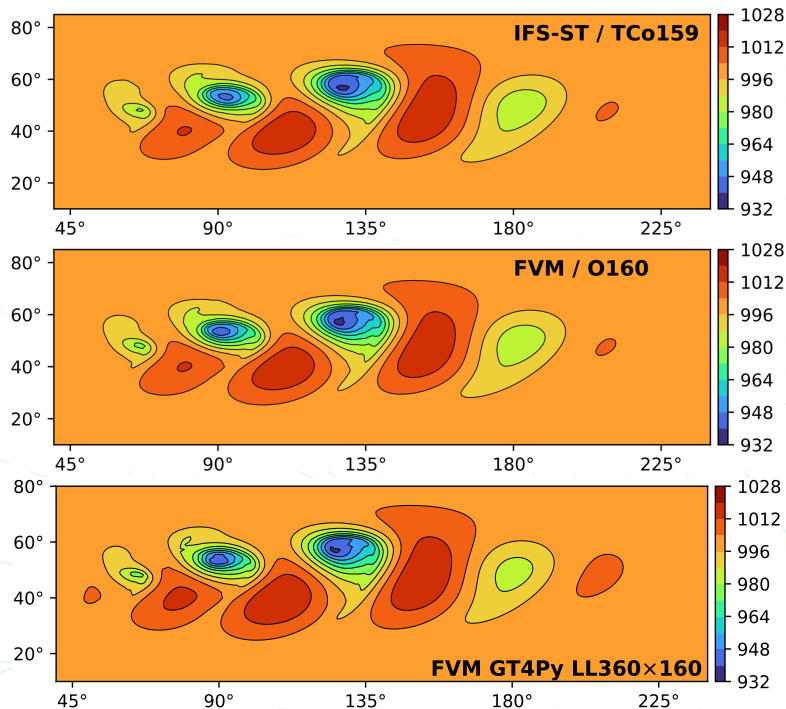


The **single model domain/user interface** in Python with embedded GT4Py for stencil computations **seamlessly drives various backends** by a simple switch, e.g.,:

- ❖ **gt:cpu_ifirst**: C++ CPU where array layout is (I, J, K)
- ❖ **gt:cpu_kfirst**: C++ CPU where array layout is (K, I, J)
- ❖ **gt:gpu**: Gridtools CUDA C++
- ❖ **cuda**: native CUDA C++
- ❖ **dace:gpu**: GPU backend leveraging the Data-Centric Parallel Programming Framework (DaCe, Ben-Hun et al. 2019, 2022). See <https://github.com/spcl/dace>.

One single user / domain interface ↔ multiple targeted backends

Validating the GT4Py based dynamical core against references models



Spectral-transform IFS (ECMWF operational dynamical core) on octahedral grid

FVM original Fortran code on octahedral grid (runs on CPUs only)

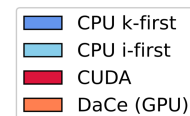
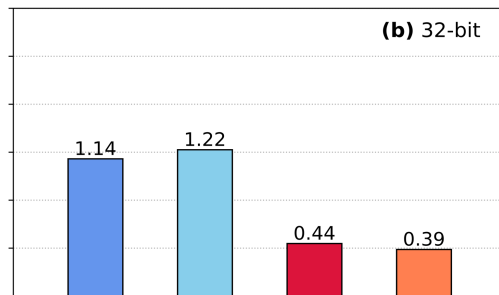
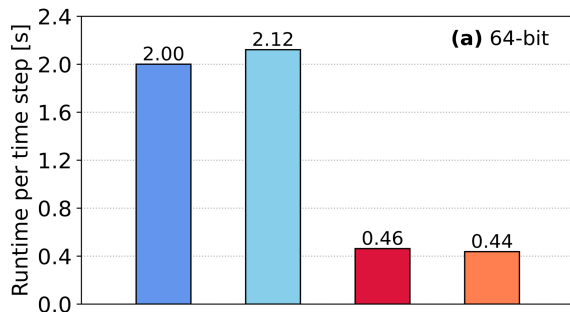
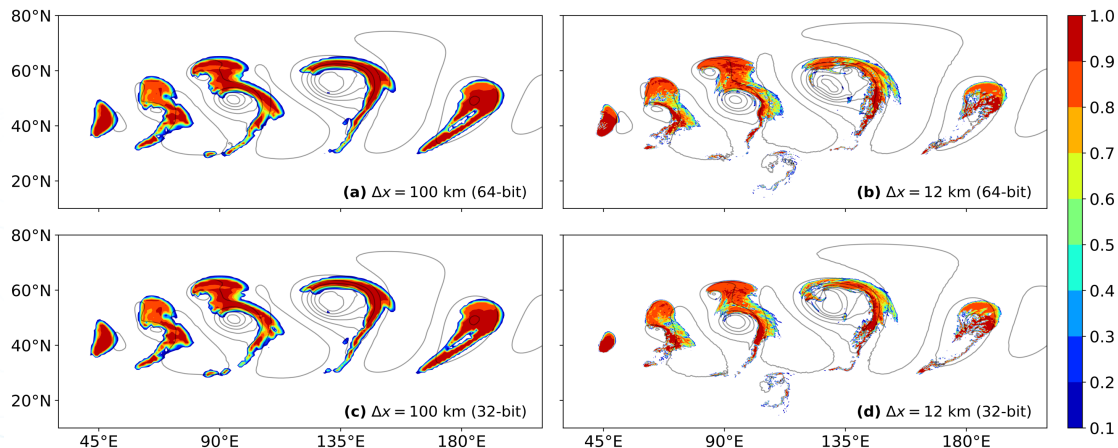
FVM implemented entirely in Python with GT4Py Version 1 on regular lat-lon grid (runs with various backends on CPUs and GPUs)

Shown is pressure on lowest level (hPa) at day 10 of DCMIP2016 baroclinic wave

Reduced precision

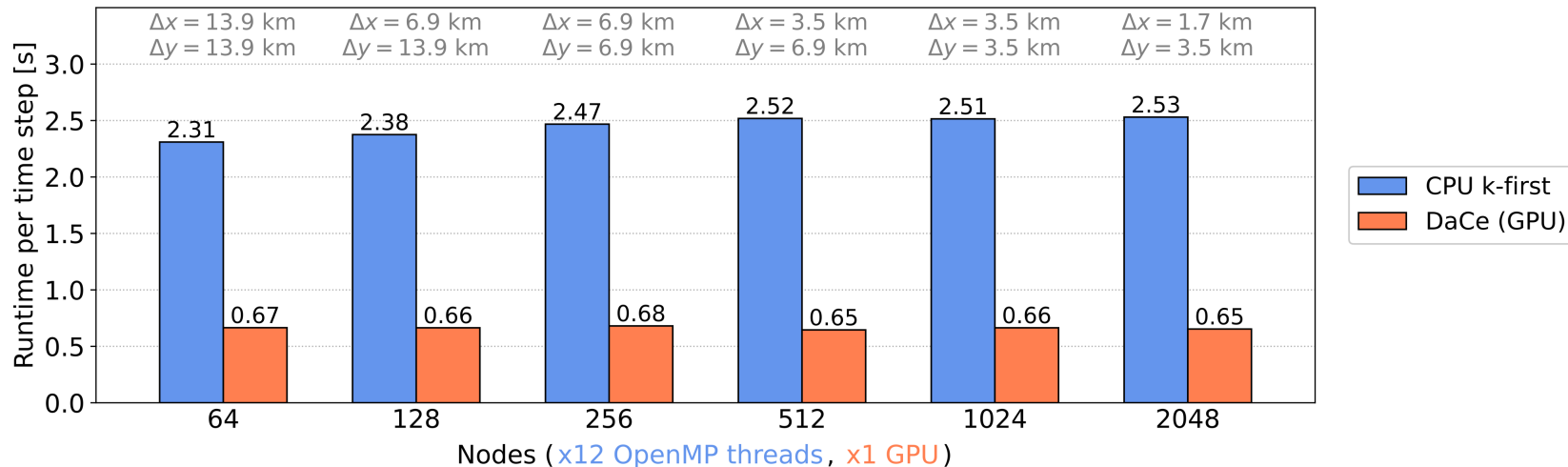
As with the original Fortran FVM code, we can run the GT4Py based FVM with either 64-bit or 32-bit precision.

Results for DCMIP2016 moist baroclinic wave with dynamical core coupled to ECMWF cloud scheme. Shown are cloud fraction at about 2 km above the surface (shading) and surface pressure (contour levels with 10 hPa interval) at day 10.



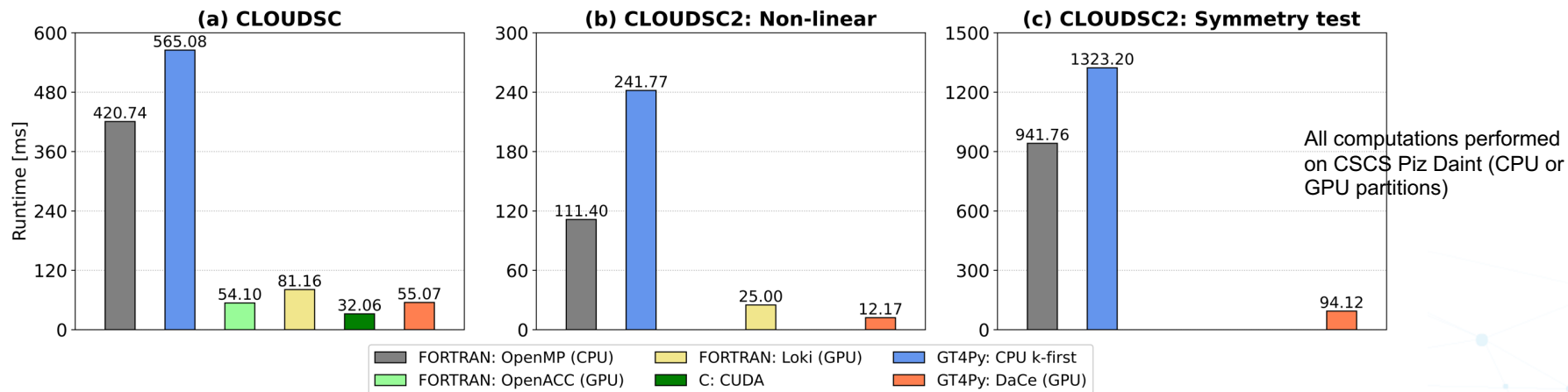
Runtime measures on CSCS Piz Daint supercomputer show increased computational performance with 32-bit precision. Expected speed-up from 64-bit to 32-bit on CPUs, but very little improvement with GPUs here.

Distributed model



- **Weak scaling of structured-grid FVM** nearly-global configuration (latitude +80deg) coupled to IFS cloud scheme.
- Here we test scaling across the CPU or the GPU partitions of CSCS' Piz Daint supercomputer.
- Halo exchanges based on GHEX -- Generic Exascale-ready Library for Halo-Exchange Operations – with Python bindings. GHEX is developed at CSCS and supported by PRACE – Partnership for Advanced Computing in Europe.

GT4Py for ECMWF parametrizations and TL/AD model code



- In the PASC funded KILOS project at ETH Zurich, we have been exploring the porting of IFS physical parametrizations to Python with GT4Py. Performance comparisons for ECMWF's prognostic cloud scheme CLOUDSC (a).
- CLOUDSC2 is the simplified cloud scheme with tangent-linear (TL) and adjoint (AD) model used in 4DVAR assimilation. Timings for running the TL/AD symmetry test on CPUs and **for the first time on GPUs** is shown in (c). Study is currently extended and prepared for publication (Ubbiali et al. in prep. 2023).

Upcoming: FVM implementation based on the declarative GT4Py

Python embedded GT4Py Field Operator
example for simple Upwind scheme:

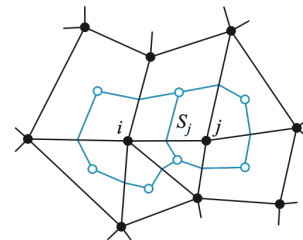
```
@field_operator
def advection_scheme_upwind(
    rho: Field[[Vertex], float],
    dt: float,
    vel: tuple[Field[[Vertex], float], Field[[Vertex], float]],
    vol: Field[[Vertex], float],
    dual_face_orientation: Field[[Vertex, V2EDim], float],
    dual_face_normal: tuple[Field[[Edge], float], Field[[Edge], float]],
    dual_face_length: Field[[Edge], float]
) -> Field[[Vertex], float]:
    flux = upwind_flux(rho, vel, dual_face_normal, dual_face_length)
    return rho - (dt / vol) * neighbor_sum(
        flux(V2E) * dual_face_orientation, axis=V2EDim)
```

Three main operator concepts:

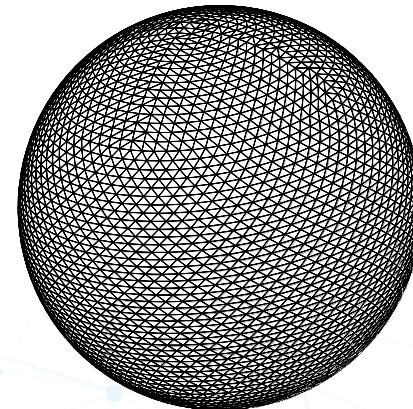
@program: Sequence of (stateful) operator calls transforming the input args and writing back the return value to a specified output field

@field_operator: covering all common patterns of finite-volume stencils with multiple field operations

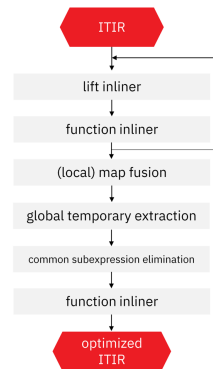
@scan_operator: expressing computations with sequential dependencies such as in direct implicit schemes and physical parametrizations



```
mesh.pole_edge_mask,
mesh.dual_face_orientation,
mesh.dual_face_normal_weighted_x,
mesh.dual_face_normal_weighted_y,
```



```
from atlas4py import StructuredGrid
grid = StructuredGrid("H18")
mesh = AtlasMesh.generate(grid)
```



Goal for 2024: Optimized declarative GT4Py-based global non-hydrostatic dynamical core in GPU distributed configuration



Thank you for listening!