

HySoP: Hybrid Simulation with Particles

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Summary

The present numerical tool HySoP is a Python package designed for high performance numerical simulations of fluid-related problems based on Remeshed Particle Methods (RPM) semi-Lagrangian methods. This tool has been successfully used for external flows, including cases involving bluff bodies, as well as for flows coupled with transport equations and flows through porous media.

In the Computational Fluid Dynamics community, the most widely used methods – such as finite difference, finite volume, finite element and spectral methods – operate within a purely Eulerian framework and have been extensively studied both in terms of consistency and stability, as well as numerical dissipation characterization.

Meanwhile, particle approaches have experienced significant development in the context of incompressible flows. These methods differ from Eulerian approaches due to their intuitive and natural description of the fluid flow, as well as their low numerical dissipation and by their ability to circumvent the nonlinearities related to the advection phenomenon. Many efforts have been devoted to addressing the main intrinsic difficulties of purely Lagrangian particle methods. These efforts have led, in particular, to the design of semi-Lagrangian approaches – also referred to as Remeshed Particle Method (RPM) – where the particles discretizing the flow are regularly remeshed onto a cartesian grid. This approach combines the strengths of both Eulerian and Lagrangian methods ([Mimeau & Mortazavi, 2021](#)).

Statement of need

The HySoP library (Hybrid Simulation with Particles) is designed for distributed hybrid architectures, supporting CPU and GPU compute devices through MPI+OpenCL. Its high level functionalities and the user interface are mainly written in Python. In general, HySoP aims to provide an architecture-agnostic, performance-portable and easily reusable numerical code. It should be noted that HySoP is a scientific research software under continuous development. Among the most advanced open-source softwares related to RPM and similar to HySoP are OpenFPM ([Incardona et al., 2019](#)) and Murphy ([Gillis & Rees, 2022](#)). Both are parallel and accelerated libraries. OpenFPM is an open-source C++ framework designed for parallel particles-only and hybrid particle-mesh codes. The interpolation schemes it provides are built to support various types of meshes, whereas HySoP relies solely on one-dimensional interpolations along the directions of a Cartesian grid. Murphy is a multiresolution adaptive grid framework for numerical simulations on 3D block-structured collocated grids designed for distributed computational architectures. This multiresolution finds its benefits dealing with unbounded

⁴² domains while HySoP is restricted to a homogeneous cartesian grid that does not require
⁴³ interpolations nor handling computational load imbalance.

⁴⁴ Features of the software

⁴⁵ Description of the software design

⁴⁶ HySoP has been designed on the basis of an decoupling between the mathematical specifications
⁴⁷ of the problem to solve and the numerical methods and algorithms. The goal is to allow the
⁴⁸ user to specify only the higher level specifications, using a formulation that remains close to
⁴⁹ the mathematical formalism:

- ⁵⁰ ■ problem parameters;
- ⁵¹ ■ computational domain;
- ⁵² ■ variables defined on domain;
- ⁵³ ■ operators;
- ⁵⁴ ■ overall discretisation for the cartesian grid.

⁵⁵ Specifying lower-level details like numerical methods, target architectures, and parallelism
⁵⁶ layout is optional. The corresponding code is interchangeable and upgradable without requiring
⁵⁷ any changes to the user's code.

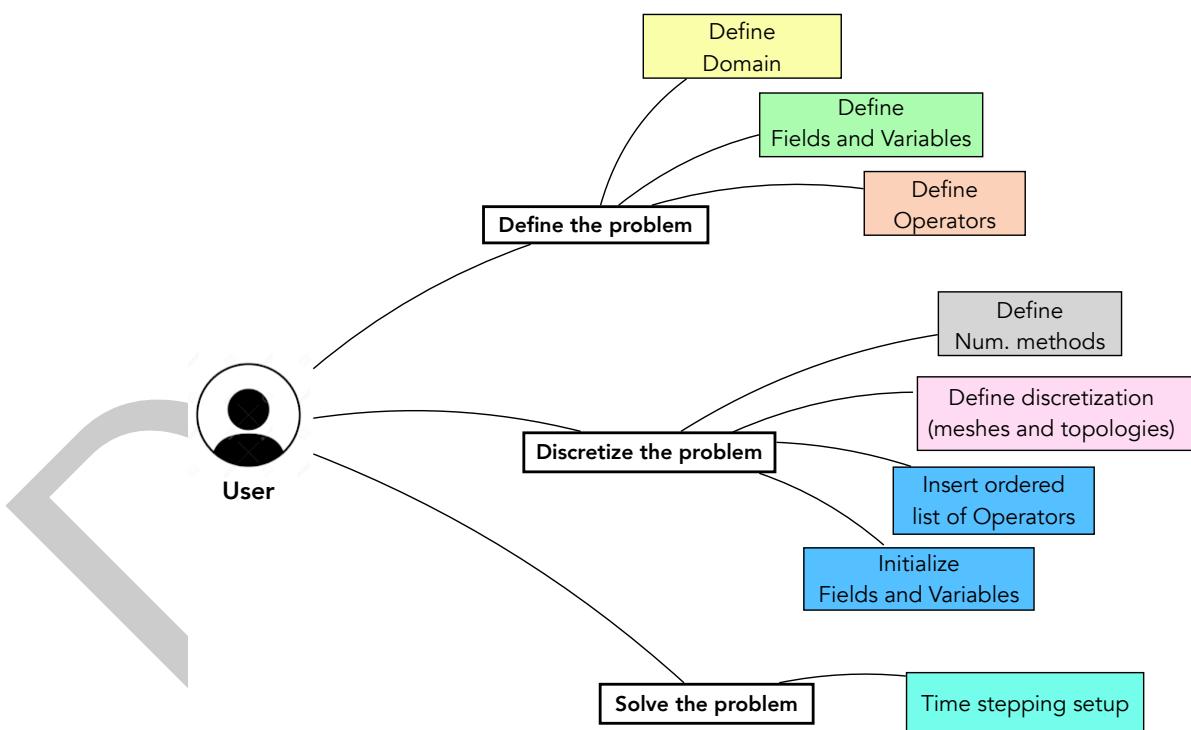


Figure 1: Use case diagram

⁵⁸ From the user's perspective, the main use case can be decomposed into three main steps (cf
⁵⁹ Figure 1):

- ⁶⁰ 1. Problem description: expressed as a mathematical PDE formalism using domain, variables
⁶¹ and operators;
- ⁶² 2. Problem initialization: after describing numerical methods and their parameters, the user
⁶³ specifies the main grid resolution, the mesh decomposition for parallel execution, and the

compute backend. Then, the user defines the operator ordering and data initialization procedures. At this stage, all memory allocations are completed, generated code is optimized and compiled, and communication layout is established.

3. Problem solving: For time-dependant problems, the user provides additional parameters (e.g. time steps) and computations proceed by sequentially applying the operators.

69 Programming languages and external dependencies

70 The main drawback of pure Python programs is performance, which is mitigated by using
71 arrays and tools from the numpy module. Additional performance improvements are achieved
72 through optimized external libraries like fftw for fast Fourier transforms. HySoP is also using
73 the f2py python module to use a Fortran implementation of RPM ((Lagaert et al., 2014)).
74 Further speedup is obtained via just-in-time compiling using numba or OpenCL towards multi-core
75 architectures. Numba translates Python code into compiled code at runtime, while OpenCL
76 requires an explicit code. In HySoP, this code is generated from formal representation of the
77 numerical methods. Additionally, micro-benchmarks are performed during initialization to
78 optimize the parameters of the generated OpenCL code.

79 The interaction between the computing backends and the base Python layer is summarized in
80 Figure 2. Our software architecture allows easy integration of new backends for which ensuring
81 complete inter-operability remains the main challenge.

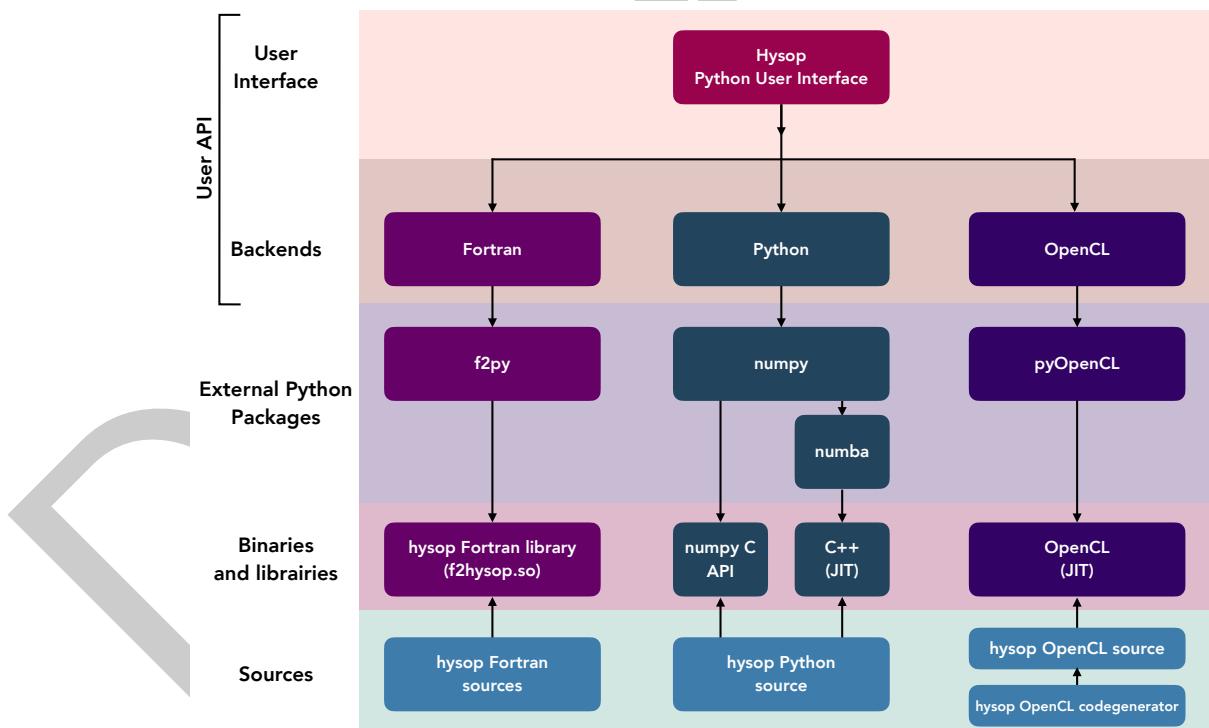


Figure 2: Computing backends in HySoP

82 Parallelism

83 The software supports distributed memory parallelism through domain decomposition relying
84 on MPI parallelism through the mpi4py interface. As a result, HySoP may be considered as a
85 Python-based solver that utilizes hybrid MPI+OpenCL programming for heterogeneous computing
86 platforms.

87 HySoP also exploits parallelism through an operator splitting approach, enabling coarse-grained

⁸⁸ task parallelism by executing simultaneously distinct and weakly coupled operators, as defined
⁸⁹ in the overall mathematical formulation.

⁹⁰ Evaluating HySoP's computational performance is challenging; however, references such as
⁹¹ (Cottet et al., 2014), (Keck, 2019), and (Keck et al., 2021) describe performance insights for
⁹² multi-GPU and heterogeneous platform simulations.

⁹³ Applications

⁹⁴ The following list illustrates the successful use of the HySoP library across a wide range
⁹⁵ of domains of applications, each involving diverse governing equations. This illustrates its
⁹⁶ versatility and flexibility:

Applications	Involved equations	Reference
- Bluff body flows	Navier-Stokes	(Mimeau et al., 2016, 2021)
- Large-Eddy Simulations (sub-grid scale modeling)	Filtered Navier-Stokes	(Crouy-Chanel et al., 2024)
- Transport of passive scalar at high Schmidt number	Navier-Stokes and a passive scalar advection-diffusion	(Cottet et al., 2014)
- Sedimentation in high Schmidt number flows	Navier-Stokes coupled with scalars advection-diffusion	(Keck et al., 2021)
- Passive flow control using porous media	Brinkman-Navier-Stokes	(Mimeau et al., 2017)
- Porous media dissolution at pore-scale	Darcy-Brinkman-Stokes coupled with reactive transport	(Etancelin et al., 2020)
- Porous media precipitation and crystallization at pore-scale	Darcy-Brinkman-Stokes coupled with reactive transport	(Perez et al., 2025)

⁹⁷ For Navier-Stokes flows HySoP relies on the operator splitting to make use of most efficient
⁹⁸ numerical methods for convection terms (RPM) and diffusion terms (Fourier solvers). Scalar
⁹⁹ transport coupled with fluid flow exploits the spatial scales separations induced by large Schmidt
¹⁰⁰ numbers using bi-level discretization on nested Cartesian grids. Porous media are handled by
¹⁰¹ the Brinkman penalization method that enables flows around permeable bodies or internal flows
¹⁰² through porous media whose possibly complex structure is difficult to capture using traditional
¹⁰³ meshes. Finally, in the context of weakly coupled sub-problems, HySoP take advantages of
¹⁰⁴ task parallelism using concurrently several architectures (e.g. flow solver on CPU and reactive
¹⁰⁵ transport on GPU).

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