PhD. Thesis

An Object Oriented and High Performance Platform for Aerothermodynamics Simulation

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Summary

This thesis presents the author’s contribution to the design and implementation of COOLFluiD, an object oriented software platform for the high performance simulation of multi-physics phenomena on unstructured grids. In this context, the final goal has been to provide a reliable tool for handling high speed aerothermodynamic applications.

To this end, we introduce a number of design techniques that have been developed in order to provide the framework with flexibility and reusability, allowing developers to easily integrate new functionalities such as arbitrary mesh-based data structures, numerical algorithms (space discretizations, time stepping schemes, linear system solvers, . . .), and physical models.

Furthermore, we describe the parallel algorithms that we have implemented in order to efficiently read/write generic computational meshes involving millions of degrees of freedom and partition them in a scalable way: benchmarks on High Performance Computing (HPC) clusters with up to 512 processors show their effective suitability for large scale computing.

Several systems of partial differential equations, characterizing flows in conditions of thermal and chemical equilibrium (with fixed and variable elemental fractions) and, particularly, nonequilibrium (multi-temperature models) have been integrated in the framework.

In order to simulate such flows, we have developed two state-of-the-art flow solvers:

1. a parallel implicit 2D/3D steady and unsteady cell-centered Finite Volume (FV) solver for arbitrary systems of Partial Differential Equation (PDE) on hybrid unstructured meshes;

2. a parallel implicit 2D/3D steady vertex-centered Residual Distribution (RD) solver for arbitrary systems of PDE’s on meshes with simplex elements (triangles and tetrahedra).

The Finite Volume (FV) code has been extended to handle all the available physical models, in regimes ranging from incompressible to hypersonic. As
far as the Residual Distribution (RD) code is concerned, the strictly conservative variant of the RD method, denominated Conservative Residual Distribution (CRD), has been applied for the first time in literature to solve high speed viscous flows in thermochemical nonequilibrium, yielding some preliminary outstanding results on a challenging double cone flow simulation. All the developments have been validated on real-life testcases of current interest in the aerospace community. A quantitative comparison with experimental measurements and/or literature has been performed whenever possible.
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Chapter 0. Acknowledgements

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Chapter 1

Introduction

In the scientific field, it often happens that a researcher or a team of researchers start writing a software package that is meant to tackle some specific problems. When new problems or new methodologies to solve problems pop up, one has to choose between adding new features into an existing code or writing a new one from scratch.

The first solution can easily lead to an unforeseen and hardly manageable growth of the original software and can result in a series of frequent changes, re-factorings and, in the worst case, full re-design.

The last solution is typically adopted when the needed changes are incompatible with the existing software or when the desired new features are far from the original goals that were planned in the long-term.

In both cases, the importance of allowing and maximizing the software reuse and modularity becomes clear. The question that arises is then: how to maximize the reusability and the flexibility of the software? To answer these needs, a great help has been given to the scientific community during the last 15 years by the Object-Oriented (OO) programming. The latter has pushed developers to improve their way of conceiving and designing software by redirecting the focus more on the way scientific concepts and algorithms interact than on the mere functionalities per se.

At the same time, the increasing power of computer architectures and the concurrent advance towards parallel and distributed computing have allowed researchers to perform more and more challenging simulations of complex physical phenomena, motivating at the same time the development of increasingly sophisticated numerical algorithms.

As an example, CFD represents a field in which powerful computer resources have become fundamental to be able to simulate complex flow fields closer to the demands of industry, medicine and, more generally, technological progress. The required power is not anymore only a matter of execution speed or memory capacity, but it has also become a matter of software design, since scientific researchers need constantly the implementation of new features, models, tools to increase the quantity and enhance the quality of
Today’s CFD applications require a considerable degree of flexibility from numerical tools, since all of the following situations can occur and should be envisioned:

- the same physical phenomena can be solved by means of different space discretization methods (Finite Volume, Finite Difference, Finite Element, Residual Distribution, Discontinuous Galerkin ...), time integrators, boundary condition treatments, iterative and direct linear system solvers etc.;
- the same numerical methods can be applied to different physical problems and their corresponding set of equations;
- some subsets of the governing equations may need to be treated with different numerical schemes;
- different physical models and constitutive relations can be applied to close the same set of governing equations.

Moreover, data structures of research codes have to be flexible and dynamic enough to support parallelization, mesh adaptation, multigrid techniques. In order to get all these capabilities in one single code, the commonly used procedural programming based on functional decomposition has proven to be insufficient. On the contrary, OO design and programming have been of great help in tackling all these rapidly increasing needs in a more appropriate and scalable way. During the last decade, this has lead to the development of several platforms for research purposes such as DiffPack [6], Deal II [12], LibMesh [70], Life [130], OpenFoam [44]), etc. and the migration of most commercial flow solvers (Fluent, StarCD, Numeca’s FINE ...) to the new approach.

1.1 The COOLFluiD project

1.1.1 History of the project

The current work has moved its first steps back in March 2002, within the COOLFluiD (Computational Object Oriented Library for Fluid Dynamics) project, aiming at creating an OO framework for the simulation of multi-physics on unstructured meshes. One of the primary goals of COOLFluiD was to port to a new common
platform the main simulation capabilities spread over a large number of unmaintained legacy codes, developed at the Von Karman Institute by the previous generation of PhD students. Such a framework was also and especially meant to be scalable, i.e. capable of evolving, maintainable and to easily accommodate new developments for many years to come, both from the physical modeling and the algorithmic sides.

The preliminary design was the fruit of the collaboration between the author and two other PhD candidates, namely Tiago Quintino (VKI) and Pieter De Ceuninck (KU Leuven).

Between January and September 2003, Quintino and the author worked on separate tracks, each one on a different implementation of the same concept design. In October of the same year, the author’s version of COOLFluiD [74] and Quintino’s HOTFluiD merge into one single platform, keeping the best capabilities and properties of both codes, but bringing them to a higher level of flexibility, thanks to a partial synergic re-design and re-implementation effort.

After the definition of a unified framework and the entry in the team of Dries Kimpe (KU Leuven) and, later on, of Thomas Wuilbaut (VKI), the scope of application was progressively enlarged to high-performance computing and more and more complex multi-physics. During the last few years, also thanks to the contributions of many other PhD candidates, diploma course members, post-docs and external partners, the project has been growing very rapidly towards the status of a customizable collaborative environment for CFD, where many researchers can cooperate to extend the simulation technology towards predefined target applications while taking maximum benefit from the others’ work.

1.1.2 The COOLFluiD Architecture

COOLFluiD consists of a collection of dynamically linked libraries and application codes, organized in a multiple layer structure, as shown in Fig. 1.1. Two kinds of libraries are identifiable: kernel libraries and modules. The former supply basic functionalities, data structure and abstract interfaces, without actually implementing any scientific algorithm, while the latter add effective simulation capabilities.

The framework was developed according to a plug-in policy, based on the self-registration [19, 74, 78] and self-configuration techniques [78, 131], which allow new components or even third party extensions to be integrated at run-time, while only the Application Programming Interface (API) has been exposed to the developers. From top to bottom of the COOLFluiD architec-
ture, one finds, progressively, more functional layers: the kernel defines the interfaces, the modules implement them and the application codes can select and load on demand the libraries needed for an actual simulation. Moreover, COOLFluiD is a component-based platform, where each functionality is enclosed in a separate component which can be connected to others at run time to create one or more actual application solver.

**Figure 1.2:** Component-based architecture of COOLFluiD: kernel modules, internal and external plug-ins.

Fig. 1.2 shows an example of some of the components available in COOLFluiD and the distinction between kernel modules, internal and external plug-ins.
is stressed. Internal plug-ins are typically developed by VKI or by partners with access to the source code, while external plug-ins are usually developed independently by third parties with access only to the API. This multi-layer and multi-component structure guarantees flexibility and modularity at the highest design level and helps managing the development process, since layers and components automatically reflect different implementation levels and, in some cases, permission rights.

1.2 Objectives of the present thesis

The original goals of the COOLFluid project have traced the guidelines for the present work, aiming at implementing a flexible and extendible object-oriented platform for high-performance simulation of aerothermodynamics on unstructured grids. This rather ambitious target has given the project a multi-disciplinary connotation nature since the start, requiring development of competence and research effort in different areas:

1. OO design tailored to scientific applications to provide the framework with a solid and flexible software infrastructure, allowing new developers to easily integrate new functionalities and fully reuse existing ones;

2. parallel algorithms and programming techniques to support large scale and high-performance computing;

3. development of robust and efficient state-of-the-art numerical solvers suitable for handling a wide range of flow regimes, ranging from incompressible to hypersonic speed;

4. integration of multiple models to provide more accurate description of complex physical phenomena, including transport properties, reaction kinetics and statistical thermodynamics.

As far as the OO design is concerned, a number of structural solutions for creating reusable and extendible scientific software have been developed in collaboration with other team members [68, 131] and will be presented. We will propose suitable design patterns for providing a uniform way of encapsulating numerical algorithms for solving Partial Differential Equation (PDE), physical models and their interaction within multi-physics simulations. We will also present solutions for handling generic mesh data-structures with a transparent treatment of both serial and distributed data.
Given our intention to support complex and computational intensive simulations, we will describe the efficient parallel algorithms for mesh decomposition and parallel input/output that we have implemented and tested on large cluster architectures.

Since our final target is to deal with aerothermodynamic applications in conditions ranging from perfect non reactive gas to thermo-chemical equilibrium to full nonequilibrium, especially in the hypersonic regime, many physical models had to be integrated in COOLFluiD and they will all be described. With regard to this, some details will be left out of this thesis, since physical modeling was not the main focus of this work and since most of the chemistry, thermodynamics and transport algorithms were used as a black box, through the interface of Mutation [94, 116, 140]. The latter is a research library which has been developed at the VKI during the last decade and has been fully linked to COOLFluiD within this work, in order to make it usable in real-life CFD simulations.

Furthermore, we will focus our attention on numerical schemes which combine accurate shock capturing and robustness. The standard cell-centered FV is the best established method and usually the preferred choice for these applications, in combination with implicit time stepping techniques to accelerate the convergence. During the last decade, however, RD schemes have been emerging as a possible valid alternative for the computation of compressible flows on unstructured meshes. We will show that our flexible platform has allowed us to develop and compare both methodologies. In particular, in this thesis we will extend, for the first time in literature, the strictly conservative variant of RD schemes, aka CRD, to the simulation of viscous chemically reactive flows.
Part I

Design Solutions for Scientific Computing
Chapter 2

Object Oriented design solutions

Introduction

Designing a framework for generic PDE solvers is not a trivial task by itself, but additional complexity has come, in our case, from having planned to deal efficiently with arbitrary numerical algorithms (with potentially different parallel data structures) and multi-physics applications. According to a simplified conceptual view, we can consider a numerical solver for PDE’s as made by a limited number of independent building blocks:

- some mesh-related data, typically numerics-dependent, that represent a discretized view in terms of both geometry and solution of the computational domain;
- an equation set describing the physical phenomena to study, typically independent of numerics;
- a group of interacting numerical algorithms to solve the PDE’s on the given discretized domain.

We can therefore identify three issues of fundamental importance for developing an environment for solving PDE’s:

- storage and handling of mesh data;
- implementation of physical models;
- implementation of numerical methods.

A good OO design should be able to tackle these three aspects orthogonally, independently one from the other as much as possible, in order to minimize coupling and maximize the reusability of the single components. To this aim, a clear separation should be enforced between physics and numerics, the first being the description of the properties, constitutive relations and
quantities that characterize the equation system, the second providing the mathematical tools (algorithms, algebraic and differential operators) to discretize and solve those equations. This section will start with a short introduction to basic concepts in object-oriented programming and design like inheritance, polymorphism and composition which will be reused heavily later on. After that, our attention will focus on analyzing each one of the three above-mentioned problems (2.2, 2.3, 2.4), and on presenting corresponding possibly effective solutions, as they were implemented in COOLFluID. Some illustrative examples will also be given in Sec. 2.3.1 and 2.4.1 in order to show the concrete applicability of the proposed ideas.

2.1 Basic concepts in OO Programming

OO programming consists in a new conception of programming based on an accurate conceptual analysis of a given problem that leads to the identification of the objects involved, their factorization into classes, the definition of class interfaces and hierarchies and the establishment of key relationships among them. This, at start, requires considerably more effort than writing a flow chart and translate it into actual code, as procedural (aka functional) programming would normally require, but it’s also considerably more powerful, because it allows software developers to go beyond the specific problem and create the basis for reusable code. OO programming is usually called programming to the interfaces, which tells a lot about the effort to generalize and abstract as much as possible to determine what are the objects involved and how these objects interact through their interfaces.

Some key ingredients of the OO approach, as indicated in [33], are

- **data abstraction**: the logical properties of data types are separated from their implementation;

- **data encapsulation**: the physical representation of the data of an algorithm is surrounded, so that the user of the data doesn’t see the implementation, but deals with the data only in terms of its logical picture, namely its abstraction;

- **information hiding**: the details of a function or data structure are hidden in order to control access to them;
• **polymorphism**: the ability to determine dynamically (at run-time) or statically (at compile time) which of several operations with the same name is appropriate.

### 2.1.1 C++ techniques

C++ has been the programming language chosen for the actual implementation of the COOLFluiD framework, since it provides a lot of attractive features and allows developers to use very effective techniques, at the cost, sometimes, of a quite complex syntax. Occasionally, some minor portability problems between one compiler and another can still be encountered, but this has become more and more rare since the standardization of the language. Another valid reason to choose C++ for implementing an OO platform nowadays is the large availability of multi-purpose libraries for scientific computing in C/C++ which are easily linkable, including the ones (LAM, MPICH, OpenMP) implementing Message Passing Interface (MPI) for parallel and distributing computing.

Some of the main OO techniques supported by C++ will now be briefly described, in order to help the reader familiarize with some terminology which will recur often in the rest of the chapter.

#### 2.1.1.1 Class Inheritance

This is a mechanism that lets developers implement one class interface defining an object in terms of another existing one. When a derived class inherits from a base class, it includes the definition of all the data and the operations defined by the parent class. A class is said to be abstract if its only purpose is to define a common interface for its subclasses. This implies that all or some implementations of the operations are deferred to the derived classes, making the abstract class non instantiable. Classes that aren’t abstract are called concrete. Fig. 2.1 shows the OMT (Object Modeling Technique) [48] notation for class inheritance, with a vertical line and a triangle pointing towards the base class.

#### 2.1.1.2 Dynamic and static binding

When identical requests that may have different implementations are forwarded to the abstract interface of a base class, inheritance can be used to obtain the already mentioned polymorphism (see 2.1). As a consequence, the correct derived class will be selected to fulfill the request appropriately.
When this selection occurs at run-time, the association is called *dynamic binding*, otherwise, when it occurs at compile time, it is called *static binding*.

### 2.1.1.3 Object Composition

Object composition is an alternative to class inheritance. It consists of assembling or *composing* objects to get more complex functionalities. The composition is defined dynamically at run-time through objects keeping references (or pointers) to other objects and forcing them to respect each others interfaces.

Each object of the composition can be considered as a *black box*, because no internal details (data, algorithms) of the objects are visible from outside the class. This allows to limit implementation dependencies and maintain encapsulation, leading to the so-called *black-box reuse* [48].

The OMT notation for composition is presented in Fig. 2.2. Composition can imply either *aggregation*, indicated by an arrow line starting with a diamond, or *acquaintance*, indicated by a plain arrow line. Aggregation requires that one object owns and/or is responsible for the other one, meaning that aggregatee and aggregator have identical lifetimes. Acquaintance implies that an object simply has knowledge of another one, without having ownership on it.
2.1.1.4 Templates

This technique allows developers to define a type (other name for class) without specifying all the other types it uses. The unspecified types are supplied as parameters at the point of use. As an example, we can consider a container (list, vector, set, etc.) that must handle different kind of elements (int, double, char, whatever fancy object) in the same way, and it’s therefore parameterized with the element type.

Templates can help to create highly generic, efficient and reusable code. The most powerful example of application of this technique is the so called Standard Template Library (STL), that has been extensively used in the present work.

For further details on how to implement this techniques in practice, which is clearly out of the scope of this work, we particularly recommend to look at [149], [98], [160], [10].

For the interested reader, however, we provide in Appendix (Sec. A.1) an example of powerful use of template-based meta-programming for achieving an optimal implementation of symbolic array algebra. In the latter, we show how to enhance the state-of-the-art Fast Expression Template (FET) technique [57, 66] in order to make it usable within software frameworks with dynamically linked plug-ins, like COOLFluiD, and in order to allow the compiler to perform an optimization known as loop unrolling.
2.2 Mesh Data Handling and Storage

PDE-driven simulations typically involve a considerable amount of data related to the discretization of the computational domain, such as coordinates, state variables, geometric entities, etc., all grouped inside container objects (arrays, lists ...) on which numerical actions are performed. Additional complexity appears when computing on unstructured grids, as in the case of COOLFluiD, since different numerical methods can require the storage and usage of different kind of connectivity information. As a result, genericity in both quantity and types of data must be addressed. Moreover, in a scalable and safe design of the data handling, framework components should be able to share data without violating encapsulation. In a high performance environment, developers of new modules should be allowed to elect newly defined data types to be stored, handled efficiently, and shared in parallel communication. Ideally, local and distributed data should be treated uniformly from the developer’s point of view, and support for different communication strategies (e.g. message passing and shared memory) should be offered.

2.2.1 Data Storage

As explained in [78, 80], in COOLFluiD, all massive data whose size scales with the problem complexity are encapsulated by a facade object MeshData, which, in particular, aggregates and offers safe access to a number of underlying instances of DataStorage.

```
1 template <class TAG> class DataStorage {};
```

**Code Listing 2.1**: Empty class DataStorage class definition

DataStorage is defined in C.L. 2.1 as a template empty class parameterized with a lightweight tag policy, TAG, which is meant to specify the actual parallel implementation or communication type, e.g. LOCAL, MPI, PVM, SHM:

```
2 class LOCAL {}; // serial communication
2 class MPI {};  // MPI-based parallel communication
2 class PVM {};  // PVM-based parallel communication
2 class SHM {};  // shared memory communication
```

**Code Listing 2.2**: Tag classes defining the communication models

As explained in [69], these tags do not appear explicitly in the code, but
they are aliased at compile time to more general concepts, such as \texttt{GLOBAL} for inter-communication or \texttt{REMOTE} for intra-communication, according to the available models and the user’s preferences.

```c
#ifdef HAVE_MPI
typedef MPI GLOBAL;
#else
typedef PVM GLOBAL;
#endif

#ifdef HAVE_SHM
typedef SHM REMOTE;
#else
typedef LOCAL REMOTE;
#endif
```

\textbf{Code Listing 2.3:} Macros definitions for \texttt{GLOBAL} and \texttt{LOCAL} communication models

In COOLFluiD, a parallel layer is defined with the task of concealing implementation details that rely upon specific parallel paradigms. To this aim, only the aliased tags are used outside the parallel layer, i.e. in the numerical modules of the framework. \texttt{DataStorage} maintains a key role within the parallel layer, by providing a uniform, user-friendly interface to create and manage arrays of generic data types, that are meant to be shared among different numerical components. In particular, \texttt{DataStorage} offers the same interface to treat both local and distributed data uniformly, even though the actual implementation of its member functions depends on the chosen \texttt{TAG} class. In other words, every choice of a \texttt{TAG} class leads to a different instantiable explicit template specialization [160] of \texttt{DataStorage}, as clarified here after.

\subsection{2.2.1.1 Local Data Storage}

The class definition of \texttt{DataStorageLocal} is reported in C.L. 2.4: it provides functions to create, get and delete local data, which are stored in a \texttt{Evector} [32]. The latter offers an improved and more efficient implementation of \texttt{std::vector} [29, 149, 150], the basic container provided by STL. In particular, \texttt{Evector} supports back insertion/deletion of elements in constant time and on-the-fly memory allocation to accommodate new entries, without any overhead. \texttt{DataStorageLocal} behaves as an archive, where ar-
arrays of data with a certain type and size are registered under a user-defined name in an associative container like a `std::map` [149] [150]. Pointers to the arrays in question are statically casted to `void*`, in order to let coexist different types in the same instance of `DataStorageLocal`. No built-in garbage collection or deletion policies based on some sort of reference counting are automatically provided by `DataStorageLocal`. This fact implies that data created with `createData()` must be consistently deleted with an explicit call to `deleteData()`: ad-hoc `setup` and `unsetup Commands` fulfill this task in each numerical module, as explained in Sec. 2.4.1.

```cpp
// --- DataStorageLocal.hh --- //

template<> class DataStorage<LOCAL> {  
  public:
    // Creates and initializes a storage
    template <class T> DataHandle<LOCAL,T> createData
       (string name, int size, T init = T());

    // Deletes a storage and frees its memory
    // Returns the number of deleted storages
    template <class T> int deleteData(string name);

    // Gets an existing storage
    template <class T> DataHandle<LOCAL,T> getData
       (string name);

  private:
    // map that stores the pointers to arrays of data
    std::map<string, void*> m_dataStorage;
};
```

**Code Listing 2.4:** Explicit specialization for LOCAL DataStorage

### 2.2.1.2 MPI Data Storage

The `DataStorage` associated to the MPI tag, namely `DataStorageMPI`, is defined in C.L 2.5. Since `DataStorageMPI` requires some MPI specific assumptions, it provides its own implementation of the same interface of `DataStorageLocal`. The main difference between `DataStorageLocal` and this class is given by the presence of a MPI `communicator`, needed by `ParVectorMPI`, which is the container type for the data to be registered in `DataStorageMPI`. 
Code Listing 2.5: Explicit specialization for MPI DataStorage

2.2.1.3 Data Handle

Once registered in either a local or MPI DataStorage, data can be accessed through a smart pointer, DataHandle, that ensures safety by preventing the data array from being accidentally deleted. It also provides inlinable accessor/mutator functions to the individual array entries and offers some additional functionalities. The class definition of DataHandle is presented in C.L. 2.6.

Code Listing 2.6: Empty definition of DataHandle

Different partial template specializations of DataHandle exist, one for each tag class. As an example, the class definition for the DataHandle corresponding to the MPI tag is given in 2.7. DataHandleMPI declares functionalities that deal with the synchronization
and handling of data belonging to the overlap region. The actual synchronization and communication is delegated to the acquainted parallel vector, \texttt{ParVectorMPI}, in which all MPI calls are encapsulated. The same functions must be provided also by \texttt{DataHandleLocal} (with an empty implementation) in order to allow the whole code to compile even without having MPI, i.e. when the \texttt{GLOBAL} tag is aliased to \texttt{LOCAL}. For sake of completeness, we report in C.L 2.8 the definition of \texttt{DataHandleInternal}, from which all \texttt{DataHandles} derive. In particular, the overloading of the subscripting \texttt{operator[]} allows to directly access array entries through the \texttt{DataHandle} interface.

### 2.2.2 Topological Region Sets

The computational domain can be considered as composed by topologically different regions which correspond to the \texttt{TopologicalRegionSet} concept in COOLFluiD. Each \texttt{TopologicalRegionSet} (TRS) consists of a list of \texttt{TopologicalRegions}, i.e. subsets of the domain ideally linked with a particular Computer Aided Design (CAD) representation. Each set of boundary surfaces on which a certain boundary condition has to be enforced can be interpreted as a different TRS, while the whole boundary can be broken into smaller patches, \texttt{TopologicalRegion} (TR), according, for instance, to the surface definition in the CAD model. If we consider, as an example, the computational domain built around an airplane, FarField, Wing, Nacelle, Fuselage can all be considered as different TRS’s. One could alternatively
choose to pack Wing, Nacelle and Fuselage together to form a unique continuous TRS, called Airplane, consisting of three TR’s. Because of consistency and convenience, not only the boundary but also the interior of the domain is assigned to one or more TRS’s, according to the needs of the chosen numerical methods (space/time discretizations, mesh adaptation, mesh movement, coupling algorithms, etc.).

The latter typically are applied on geometric entities (cells, faces, edges), which represent the basic entities in the topological discretization of the domain. A **GeometricEntity** is defined by an algorithm-dependent agglomeration of degrees of freedom, both in geometric (Nodes or coordinates) and solution space (States or solution vectors). Both Nodes and States are provided with global (inter-process) and local (intra-process) IDs, which are used to build connectivity information. As a result, each TRS’s (or TR’s) can be logically viewed as set (or subset) of GeometricEntity (GE) in which each portion of the domain can be decomposed for computational purposes.

A TRS is a **Proxy** object [48] that controls the access to:

- the local and global IDs of all the GE’s belonging to the TRS;

---

```cpp
// --- DataHandleInternal.hh --- //

template<
    template <class T> class STYPE, class T>
class DataHandleInternal {
public:
    typedef STYPE<T> StorageType;

    // constructor
    DataHandleInternal(StorageType *const ptr) : m_ptr(ptr) {}
    // copy constructors, assignment operator ...

    // overloading of subscripting operator
    T& operator[](int idx) { return (*m_ptr)[idx]; }

    // overloading of operator ()
    T& operator() (int i, int j, int stride)
    { return (*this)[i*stride+j]; }

    // ...
protected:
    StorageType* m_ptr; // pointer to the array to be handled
};
```

**Code Listing 2.8:** DataHandleInternal interface
• the local IDs of all nodes and states referenced by local GE’s;
• the geo-type information (i.e. the unique IDs corresponding to the shape and polynomial order of each GE);
• the geo-to-node and geo-to-state connectivities;

as shown in the diagram in Fig.2.3. In order to minimize the memory requirements, all these data are stored sequentially in unidimensional arrays (but logically bidimensional) which can be addressed by the client code exclusively through the TRS or TR interfaces, with the latter offering only a partial view of the whole data.

2.2.2.1 On-the-fly creation of Geometric Entities

As mentioned in the previous paragraph, numerical algorithms are applied on cells, faces, edges, all corresponding to GE instances in our design. A GE is a locally/globally indexed object which is composed by States, Nodes, ShapeFunctions for the solution and geometric interpolation and it can keep reference to other neighbor GE’s, according to the computational stencil.
needed by the numerical method. A simplified interface for GE is shown in C.L. 2.9.

```cpp
// --- GeometricEntity.hh --- //

class GeometricEntity {
    public:
        // constructor and destructor, utility functions ...

        // accessor functions ...
        SafePtr<vector<State*> > getStates() const;
        SafePtr<vector<Node*> > getNodes() const;
        SafePtr<vector<GeometricEntity*> > getNeighborGeo() const;

        // mutator functions
        void setState(int is, State* state);
        void setNode(int in, Node* node);
        void setNeighborGeo(int ig, GeometricEntity* neighborGeos);

    protected: //data
        int m_geoID; // local ID of this GeometricEntity
        std::vector<State*> m_states; // state list
        std::vector<Node*> m_nodes; // node list
        std::vector<GeometricEntity*> m_neighbors; // neighbor GEs
        ShapeFunction* m_solSF; // solution shape function
        ShapeFunction* m_geoSF; // geometric shape function
};
```

**Code Listing 2.9:** GeometricEntity interface

As one can easily notice, a `GeometricEntity` is a relatively lightweight object, but it would be prohibitively expensive to carry in memory all the cells and/or faces and/or edges corresponding to the given mesh and numerical algorithm along the whole simulation. It is however extremely convenient to have such an object that encapsulates all the needed local connectivity, shape function-related, stencil-related data and that offers them ready-to-use.

This lead us to the idea of applying a *Builder* pattern [48] to separate the construction of GE’s from their representation, starting from a preallocated memory pool. The pattern is represented in Fig. 2.4.

We define a generic interface for `GeoEntityBuilder`, which is a template class parameterized with the actual builder, called `GEOBUILDER` in C.L. 2.10. `GeoEntityBuilder` just forwards actions to the aggregated `GEOBUILDER`, which is imposed by the chosen numerical algorithm.

Each possible choice for the parameter `GEOBUILDER` corresponds to a dif-
Figure 2.4: Builder pattern for creating GeometricEntities.

```cpp
// --- GeoEntityBuilder.hh --- //

3 template <class GEOBUILD> 
   class GeoEntityBuilder { 
      public: 
         // constructor, destructor 

            // allocate the prototype GEs and set up GEOBUILD data 
         void setup() {m_builder.setup();}

            // get the input data for the GEOBUILD 
      typename GEOBUILD::GeoData& getDataGE() 
         {return m_builder.getDataGE();}

         // build the GE corresponding to the given input data
         GeometricEntity* buildGE() {return m_builder.buildGE();}

         // release the built GEs to prepare the next creation
         void releaseGE() {m_builder.releaseGE();}

   private: //data 
      GEOBUILD m_builder; // actual GE builder 
   };

Code Listing 2.10: GeoEntityBuilder interface
different instance of GeoEntityBuilder. As an example, C.L. 2.11 shows the class definition of a simple FEMCellBuilder. The latter, in particular, provides a nested struct GeoData, holding the numerics-dependent input data for the creation of a GE.

During the setup phase of the simulation, a small number of GE objects, corresponding to all the entities belonging to the computational stencil required by the actual application, is created and stored in an array. While applying numerical algorithms, after the GeoData are set by the statically bound client code, the builder is asked to assemble a ready-to-use GE, where all the data have been fetched out from the corresponding TRS and DataHandles. The builder operates on the prototype GE’s that have been already previously allocated in memory, but it changes their content (data pointers) according
to the current request. The GE creational algorithm can be as complex as one needs, but all its complexity and data dependencies are hidden to the client code that, in our example, just reduces to C.L. 2.12.

```
SafePtr<GeometricEntityPool<FEMCellBuilder>> builder;
builder.setup();
...
FEMCellBuilder::GeoData& geoData = builder.getDataGE();
geoData.trs = getCurrentTRS();
geoData.idx = getCurrentGeoIDInTRS();
builder.buildGE();
...
builder.releaseGE();
```

Code Listing 2.12: FEMBuilder interface

This separation between the GE construction and the GE data object itself has allowed us to easily deal with the most diverse space discretizations and associated data structures such as Finite Element Method (FEM), cell centered FV, Spectral FV, Spectral FD, DG, each one providing an ad-hoc definition and implementation of a concrete GE builder. GeoEntityBuilder offers an example of static polymorphism, where, instead of defining an abstract interface and implement its virtual member functions in the derived classes, actions are delegated by GeoEntityBuilder to an aggregated interface which is statically bound to the client code. This choice offers superior performance (no virtual calls on some inlinable functions in tight loops) and it reflects a logical dependence of the concrete numerical method on the concrete way of creating suitable GE’s for it.

Memory optimization. On-the-fly creation of GE’s has allowed us to heavily optimize the code in terms of memory, compared to a previous approach where all GE’s were allocated and actually stored in the TRS’s. This improvement made memory usage competitive against procedural (as opposite to object-oriented) highly optimized numerical codes, such as Elsyca’s PlatingMaster, an industrial FEM solver (written in C) for electrochemical applications.
2.3 Implementation of Physical Models

Scientists and researchers that work in the field of Fluid Dynamics deal regularly with complex systems of convection-diffusion-reaction PDE’s that can be formulated as

\[
\frac{\partial U}{\partial t} + \nabla \cdot F^c = \nabla \cdot F^d + S \tag{2.1}
\]

where \( U \) (state vector of unknown variables), \( F^c \) (convective fluxes), \( F^d \) (diffusive fluxes), \( S \) (source or reaction term) depend on the chosen physical model. The latter can be seen as a composition of entities, e.g., transport coefficients, thermodynamic properties and quantities that contribute to define the mathematical description of some physical phenomena. It can be noticed that one can look at the same physical model through different formulations of the corresponding equations, which involve the use of different sets of variables, transformations, or other adaptations tailored towards a specific numerical method.

Moreover, in a framework for solving PDE’s which supports different numerical techniques and run-time loading of components, the modules where algorithms are implemented should be as independent as possible from the modules dealing with the physics description. Therefore, when new physical models or numerical components are integrated, the need for modifications on the kernel, where most of the key abstractions are defined, should be avoided. Ideally, it should be possible to build more complex models by simply extending and reusing available ones, or solving the same equations by means of different numerical techniques, without requiring changes in the existing abstract interfaces and without establishing a high level direct coupling between physics and algorithms. Widely used OO CFD platforms like [6] or [44] lack these last features, since the implementation of the physics is directly tangled to the numerical solver, in particular to the space discretization. This is surely a reasonable down-to-earth solution, it is probably efficient, but it does not promote enough reusability and interchangeability between physics and numerics. The Perspective pattern, that will now be described, proposes a structural way for overcoming the above mentioned pitfalls and for facilitating dynamical incremental changes and code reuse.

2.3.1 Perspective Pattern

Trying to define a single abstract interface for a generic physical model would be quite a demanding task and it would easily lead to a non maintainable
solution, especially if we keep in mind our concern with run-time flexibility. In COOLFluiD, this hypothetical hard-to-define interface is therefore broken into a limited granularity of independent Perspective objects [78, 80], each one offering a different view of the same physics, according to the needs of possibly different numerical algorithms. To make an example, convective, diffusive, inertial, reaction or source terms of the equations can all be Perspectives with a certain abstract interface. Moreover, if one implements a new numerical scheme that needs to access an available physical model, but that requires something not foreseen a priori and, therefore, not offered by the available interfaces, a new Perspective can be created, without affecting the existing ones.

Figure 2.5 shows the OMT class diagram of a Perspective pattern applied to a generic physical model. The base PhysicalModel defines a very general and narrow abstract interface. ConcreteModel derives from PhysicalModel, implements the virtual methods of the parent class and defines another interface to which the concrete Perspective objects (and only those) are statically bound. This other interface offers data and functionalities which are typical of a certain physics, but invariant to all its possible Perspectives.

![Figure 2.5: Perspective pattern applied to a Physical Model.](image)

The resulting pattern lets the numerical client code make use of the physi-
cal model through an abstract layer, dynamically enlargeable if required, given by a number of Perspective objects, while all their collaborations with the ConcreteModel are completely hidden. The pattern reflects the composition-based Adapter described in [48], but with a single shared Adaptee object, i.e. ConcreteModel, and multiple abstract Targets (called Perspectives here), each one with a number of derived classes (Adapters). The fact that several objects may be defined to describe the same physics may look like a disadvantage, but, in our experience, improves reusability, allows developers to easily decouple physics from numerics and gives better support to a run-time plug-in policy.

Unlike in other OO CFD platforms [28, 44] where no neat distinction between numerical algorithms and physical description is searched and where a close form of reliance (inheritance) binds the equation model and the scheme, in COOLFluID, physics is generally completely independent and unaware of numerical methods (space and time discretizations, linear system solving, etc.). However, a design that is based on Perspective objects does not even prevent from having numerical objects, such as Strategies or Commands in 2.4.1, with static binding to the actual physics, e.g. in the case of some special schemes or boundary conditions. In the latter case, the use of Perspectives can still help to limit dependencies, improve reusability and avoid excessive sub-classing.

Some code samples are now presented in order to show the concrete applicability of the described pattern for handling complex physical models.

2.3.1.1 Physical Model Object

At first, the interface of the base PhysicalModel is defined in C.L. 2.13. A polymorphic BaseTerm object, as defined in C.L. 2.14, and a corresponding enumerated variable are assigned to each independent physical term of the equation (convection, diffusion, reaction, dispersion, inertia, etc.) and they are registered in an associative container, such as a std::map. BaseTerm manages a number of unidimensional arrays of floats (RealVector), where some physics-dependent data (thermodynamic quantities, transport coefficients, parameters useful for calculating fluxes, eigenvectors, eigenvalues etc.) are stored and continuously recomputed during the simulation.

In order not to affect the total memory requirements of the computation, the maximum number of these arrays is determined by the numerical method and scales with the number of degrees of freedom in one or more geometric entities (cells, faces, etc.), typically depending on the required computational stencil. To make an example, for a cell-vertex based algorithm
(FEM, RD, etc.), the total number of physical data arrays which are stored in BaseTerm could be the maximum number of quadrature points in a cell. The size and the content of these arrays will be defined by the classes deriving from BaseTerm.

A template compositor object is defined for each possible combination of physical equation terms (Convection, Diffusion, ConvectionDiffusion, etc.). It derives from PhysicalModel and aggregates generic policies [10], i.e. subclasses of BaseTerms. In C.L. 2.15 we present the class definition of a ConvectionDiffusion compositor, while its constructor implementation is shown in C.L. 2.16. The equation terms aggregated by the compositor are registered in the parent class and each one of them is made accessible polymorphically through the getTerm() method defined in PhysicalModel.

**Example: Navier-Stokes Model.** We consider the case of a Navier-Stokes model: an overview of its class diagram is provided in Fig. 2.6.

The class definitions for the convective EulerTerm and the diffusive NSTerm
// --- BaseTerm.hh --- //

class BaseTerm {
public:
    // constructor, virtual destructor ...

    // sets the maximum number of physical data arrays:
    // this size must be set by the numerical solver
    // because it depends on the computational stencil
    void setNbDataArrays(int maxSize);

    // get the size of each array of temporary physical data
    virtual int getDataSize() const = 0;

    // finds the array of data corresponding to the given
    // degree of freedom (state vector)
    SafePtr<RealVector> getPhysicalData(State* state);
};

Code Listing 2.14: BaseTerm interface

// --- ConvectionDiffusion.hh --- //

template <class CT, class DT>
class ConvectionDiffusion : public PhysicalModel {
public:
    // constructor, virtual destructor,
    // overridden parent class pure virtual methods
    protected:
    std::auto_ptr<CT> m_convTerm; // convective term
    std::auto_ptr<DT> m_diffTerm; // diffusive term
};

Code Listing 2.15: ConvectionDiffusion interface

template <class CT, class DT>
ConvectionDiffusion<CT,DT>::ConvectionDiffusion(string name) :
    PhysicalModel(name),
    m_convTerm(new CT(CT::getTermName())),
    m_diffTerm(new DT(DT::getTermName()))
{
    // register all the equation terms and their TypeIDs
    registerTerm(PhysicalModel::CONVECTION, m_convTerm.get());
    registerTerm(PhysicalModel::DIFFUSION, m_diffTerm.get());
}

Code Listing 2.16: ConvectionDiffusion constructor implementation
Figure 2.6: Perspective pattern applied to a Navier-Stokes model.

are shown in C.L 2.17 and 2.18.

```
// --- EulerTerm.hh --- //

class EulerTerm : public BaseTerm {
public:

    // enumerated variables: density, pressure, enthalpy, etc.
    enum {RH0=0, P=1, H=2, E=3, A=4, T=5, V=6, VX=7, VY=8, VZ=9};

    // number of the enumerated variables
    virtual int getDataSize() const {return 10;}  

    CFreal getMachInf() const; // freestream mach
    CFreal getTempRef() const; // freestream temperature
    CFreal getPressRef() const; // freestream pressure

    static string getTermName() {return "Euler";}  

    SafePtr<TDLibrary> getThermodynamicLibrary() const;
private:
    ... // member data
};
```

Code Listing 2.17: EulerTerm interface
```cpp
// --- NSTerm.hh --- //

class NSTerm : public BaseTerm {
public:
  // dynamic viscosity, thermal conductivity
  enum {MU=0, LAMBDA=1};

  // number of the enumerated variables
  virtual int getDataSize() const { return 2; }

  CFreal getPrandtl() const;   // Prandtl number
  CFreal getReynoldsInf() const;  // Reynolds number
  static string getTermName() { return "NS"; }

  // get library to compute transport properties
  SafePtr<TPLibrary> getTransportLibrary() const;
private:
  ... // member data
};
```

**Code Listing 2.18: NSTerm interface**

Besides providing accessors methods for useful parameters and libraries computing physical quantities (i.e. thermodynamics or transport properties), `EulerTerm` and `NSTerm` define the size and, by means of enumerated variables, the entries for the arrays of physical data that are stored in `BaseTerm` and that have been previously introduced. An instantiable `NavierStokesModel`, i.e. what is called `ConcreteModel` in the Perspective pattern, can now be defined as in C.L. 2.19.

```cpp
// --- NavierStokesModel.hh --- //

class NavierStokesModel : public ConvectionDiffusion
  <EulerTerm, NSTerm> {
public:
  // constructor, virtual destructor
  virtual int getDimension() const;    // geometric dimension
  virtual int getNbEquations() const;  // number of equations
  virtual int setup() const;          // set up data
};
```

**Code Listing 2.19: NavierStokesModel interface**

`NavierStokesModel` can exploit the knowledge of both its concrete convective and diffusive terms to compute, for instance, adimensionalizing coefficients.
2.3.1.2 Variable Sets

Numerical algorithms are not direct clients of the ConcreteModel. As explained previously, the latter is used through a layer of polymorphic Perspective objects, whose collaborations are statically bound to the ConcreteModel and therefore potentially very efficient. A variable set, (VarSet), is a Perspective that decouples the usage of a physical model from the knowledge of the used variables. This, for instance, allows the client code to solve equations formulated in conservative variables, to perform intermediate algorithmic steps and to update in other ones (primitive, characteristic, etc.). In order to deal with a system of convection-diffusion equations, two variable sets classes are defined: ConvectiveVarSet and DiffusiveVarSet, respectively in C.L. 2.20 and 2.21.

```cpp
// --- ConvectiveVarSet.hh --- //

class ConvectiveVarSet {
public:
    // constructor, virtual destructor, etc.
    virtual void setup()=0; // set up private data

    // set physical data starting from a state vector
    virtual void setPhysicalData
    (const State& state, RealVector& data)=0;

    // set state vector starting from physical data
    virtual void setFromPhysicalData
    (const RealVector& data, State& state)=0;

    virtual void setJacobians(...)=0; // set jacobians matrices
    virtual void splitJacob(...)=0; // split jacobian
    virtual void computeEigenSystem(...)=0; // set
    // eigenvalues/vectors
    virtual void computeEigenValues(...)=0; // set eigenvalues
    virtual void computeFlux(...)=0; // compute the convective flux
};
```

**Code Listing 2.20**: ConvectiveVarSet interface

Among all the methods declared in the above two classes, particular attention is due to `setPhysicalData()` and its dual `setFromPhysicalData()`: the former starts from a given state vector (unknown variables) and sets the physical dependent data, whose entry pattern is defined by the the subclasses of `BaseTerm`; the latter does the opposite. In the case of the `EulerTerm`, for
instance, if we assume that a State $P$ holds primitive variables (pressure, velocity components, temperature)

$$P = [p, v_x, v_y, v_z, T]$$ \hspace{1cm} (2.2)

setPhysicalData() will compute an array $W$ of physical quantities (density, pressure, total enthalpy, total energy, sound speed, temperature, velocity module and components):

$$W = [\rho, p, H, E, a, T, V, v_x, v_y, v_z]$$ \hspace{1cm} (2.3)

from $P$ by means of thermodynamic relations. $W$ will be then reused to compute eigenvalues, fluxes etc. A key point for the flexibility of the design is that each VarSet has acquaintance of the physical equation terms, but not of the resulting ConcreteModel: this allows the developer to compose progressively more complex models with full reuse of the individual equation terms.

**Example: Navier-Stokes VarSets.** As shown schematically in Fig 2.6 and, in actual code, in C.L. 2.22 and 2.23, EulerVarSet and NavierStokesVarSet inherit the interface of the corresponding parent classes. Some functionalities associated to a VarSet are independent on the variables in which one needs to work: advective fluxes, for instance, can always
be computed once that physical data like pressure, enthalpy, velocity are known, because their formulation is always the same due to the conservation property.

However, as a counter example, the Euler equations can be written in conservative, symmetrizing, entropy, characteristic (...) variables and each one of this formulation defines different jacobian matrices for the convective fluxes. The VarSet abstraction is particularly useful in these more complex cases, since we can let variable-dependent subclasses of EulerVarSet implement the jacobian matrices, according to the chosen formulation. Therefore you can have EulerCons (conservative), EulerPrim (primitive), EulerChar (characteristic), etc.

It is helpful to have derived classes for the NavierStokesVarSet too, since, for instance, the transport properties can be computed differently in chemically reactive or non reactive flows, in Local Thermodynamic Equilibrium (LTE) or in Thermo-Chemical Nonequilibrium (TCNEQ).

The method computeFlux() that calculates the diffusive flux is a template method [48] in NavierStokesVarSet and delegates the computation of the transport properties to subclasses, which have more specific knowledge. In the case of the "basic" Navier-Stokes model, as indicated in Fig. 2.6, we can

```cpp
// --- EulerVarSet.hh --- //

class EulerVarSet : public ConvectiveVarSet {
public:
  // constructor, virtual destructor,
  // overridden parent virtual methods
protected:
  SafePtr<EulerTerm> m_model;
};

Code Listing 2.22: EulerVarSet interface

// --- NavierStokesVarSet.hh --- //

class NavierStokesVarSet : public DiffusiveVarSet {
public:
  // constructor, virtual destructor,
  // overridden parent virtual methods
protected:
  SafePtr<NSTerm> m_model;
};

Code Listing 2.23: NavierStokesVarSet interface
```
have `NavierStokesCons` for conservative variables and `NavierStokesPrim` for primitive variables.

### 2.3.1.3 MultiScalarTerm and MultiScalarVarSet

The Perspective pattern is especially designed to deal with cases where additional sets of equations must be attached to an existing equation system. In all these situations, a general treatment is applied, based on the combined use of `MultiScalarTerm` and `MultiScalarVarSet`. The former implements a generic extension to an existing convective physical term (e.g. `EulerTerm`), allowing to add an arbitrary number of subsets of $N_l$ physical data (scalar variables), each one corresponding to the $w_l$ quantity in a generic conservation equation written in the form:

$$
\frac{\partial \rho w_l}{\partial t} + \nabla \cdot (\rho w_l \mathbf{v}) + \cdots = 0, \quad l = 1, \ldots, N_l
$$

(2.4)

```cpp
// --- MultiScalarTerm.hh --- //

template <typename BASE>
class MultiScalarTerm : public BASE {
  public:
    // constructor, virtual destructor, setup functions
    
    // number of the enumerated variables
    virtual int getDataSize() const
    { return BASE::getDataSize() + m_nbScalarVars.sum(); }

    // start position of scalar vars data of subset i
    int getFirstScalarVar(int i) const { return m_firstVars[i]; }

    // number of scalar vars data of subset i
    int getNbScalarVars(int i) const { return m_nbScalarVars[i]; }

    // number of subsets of scalar vars
    int getNbScalarVarSets() const { return m_nbScalarVars.size(); }

  private:
    std::valarray<int> m_nbScalarVars; // number of scalar vars
    std::valarray<int> m_firstVars; // ID of the first vars
};
```

**Code Listing 2.24: MultiScalarTerm interface**

To make a concrete example, in the case of a thermo-chemical non equilibrium model, 2 subsets are needed: one corresponds to the $N_s$ continuity
equations for the chemical species, the other one corresponds to the $N_v$ conservation equations for the vibrational energy of molecules. The interface of MultiScalarTerm is shown in C.L 2.24. Analogously, MultiScalarVarSet, whose interface is shown in C.L. 2.25 allows to treat the convective part of additional equations like 2.4 in concrete ConvectiveVarSets. This allows the developer to incrementally build more and more complex physical models, where new equations can easily be added.

```
1 // --- MultiScalarVarSet.hh --- //

template <class BASEVS>
4 class MultiScalarVarSet : public BASEVS {
public:
7   typedef MultiScalarTerm<typename BASEVS::PTERM> PTERM;

10   // constructor, virtual destructor
11   ~MultiScalarVarSet() { delete m_mScalarModel; }<br>

13   // return the convective term
16   SafePtr<PTERM> getEqModel() const { return m_mScalarModel; }

19   static std::vector<EqSetData>& getEqSetData() {
22   static std::vector<Framework::EqSetData> eqSetData;
25   return eqSetData;
28   protected:
31   // add an equation data with the specified number of equations
34   void addEqSetData(int nbEqs);

37   // compute eigenvalues and convective fluxes for scalar variables
40   virtual void computeEigenValues(...);
43   virtual void computeFlux(...);
46   private:
49   SafePtr<PTERM> m_mScalarModel; // convective term
};
```

**Code Listing 2.25:** MultiScalarVarSet interface

In particular, one **EqSetData** object is associated to each equation subset in order to keep knowledge of the corresponding subset ID and provide
the list of all the variables IDs belonging to that subset.

### 2.3.1.4 Variable Transformers

Another polymorphic Perspective is the `VariableTransformer`. This allows the client code to apply matrix similarity transformations with the form $\partial U/\partial V$ between variables or to compute one set of variables from another $U(V)$ analytically (e.g. Euler primitive from Euler conservative and vice versa).

The combined use of `VarSets` and `VariableTransformers` gives the freedom to use different variables to update the solution, compute the residual, linearize jacobians, distribute residuals, without requiring any modification in the client numerical algorithms that work with dynamically bound Perspective objects, completely unaware of the actual physics.

The support for variable manipulation and variable independent algorithms which is offered by COOLFluiD represents an important feature that none of other well known CFD platforms such as [6] or [44] have.

### 2.4 Implementation of Numerical Methods

The solution of PDE’s requires the implementation of different numerical techniques that deal with time discretization, space discretization, linear system solving, mesh adaptation algorithms, error estimation, mesh generation, etc.

In a typical OO design, e.g. like the ones proposed by [28], [43], [104], or [134], each one of these simulation steps is enclosed in separate object (or polymorphic hierarchies of objects) and different patterns are applied to make them all stick together and interact.

It would be advantageous to have just a single pattern, extraordinarily reusable, that allows the developer to encapsulate different numerical methods uniformly, and lets him/her focus more on the algorithm itself rather than on its already well-defined surrounding. This pattern should ease the integration of new algorithms, but also the implementation of different versions or parts of the same algorithm, while looking for optimal solutions and/or tuning for run-time performance.

One of the main achievements of COOLFluiD, as opposed to other similar OO environments, lies in having developed a uniform high level structural solution, potentially able to encapsulate and tackle efficiently, we believe, all possible numerical algorithms: the compound Method-Command-Strategy
(MCS) pattern.

### 2.4.1 Method-Command-Strategy Pattern

The MCS pattern [78, 80, 132] provides a uniform way to implement numerical algorithms and to compose them according to the specific needs.

![OMT diagram of the Method-Command-Strategy pattern](image)

**Figure 2.7:** OMT diagram of the Method-Command-Strategy pattern.

As sketched in Fig. 2.7, **BaseMethod** defines an abstract interface for a specific type of algorithm (e.g. LinearSystemSolver, ConvergenceMethod, MeshCreator, SpaceMethod, ErrorEstimator ...). **ConcreteMethod** implements the virtual functions of the corresponding parent class by delegating tasks to ad-hoc **Commands**, see [10] [48], that share **ConcreteMethodData**, a tuple aggregating multiple polymorphic receivers (**Strategies**) [48].

Three levels of abstraction and flexibility can be identified in this pattern: **BaseMethod, Command** and **Strategies** can all be overridden, allowing one to implement the same task, at the corresponding level, in different ways. This kind of behavioral modularity allows the developers to easily re-implement or tune components (**Methods, Commands** or **Strategies**), and gives them the freedom to move code from one layer to another, according to convenience, taste or profiling-driven indications. The fast-path code, critical for the overall performance, can be wrapped inside **Commands** or **Strategies**
and it can be substituted with more efficient implementations without implying changes in the upper layer. Moreover, while freedom is left to define the abstract interface of a new polymorphic BaseMethod or Strategy, the interface of a Command is frozen, it defines only three actions, as shown in C.L. 2.26.

```cpp
// --- Command.hh --- //
class Command {
  public:
    // constructor, destructor
    virtual void setup() = 0; // setup private data
    virtual void unsetup() = 0; // unsetup private data
    virtual void execute() = 0; // execute the action
}
```

**Code Listing 2.26: Command class definition**

In COOLFluiD, managing the collaboration between different numerical methods is eased by the fact that Commands can create their own local or distributed data and share them with other Commands defined within other Methods, by making use of the DataStorage facility, whose details are presented in section 2.2.1. Moreover, Perspective objects, as described in 2.3.1, can be used within the MCS pattern, at the Strategy level, as part of the ConcreteMethodData, in order to bind a numerical algorithm to the physics polymorphically. This collaboration between MCS and Perspective patterns is an effective solution that decouples physics and numerics as much as possible. It makes possible, for instance, to employ a certain space method to discretize the equations related to different physical models, but also to apply different space methods, requiring completely different data-structures to discretize the same equations.

Interchangeability of Methods, Commands, Strategies can be facilitated and maximized by making them self-registering and self-configurable objects [19] [78, 80]. These last two concepts will be explained in Sec. 2.5. A deeper analysis of the MCS pattern and more implementation details are provided in [131, 132]. Among the possible BaseMethods that can be identified in a CFD simulation, and among those we have actually implemented in COOLFluiD, we consider now a few examples in order to show the concrete applicability of the MCS pattern.
2.4.1.1 MCS Example: SpaceMethod

In C.L. 2.27, we define the interface of a SpaceMethod, that takes care of the spatial discretization of the given set of PDE, according to a specified numerical scheme and on a chosen mesh.

```cpp
// --- SpaceMethod.hh --- //
class SpaceMethod : public Method {
  public:
    SpaceMethod(string name); // constructor
    virtual ~SpaceMethod();   // virtual destructor
    virtual void setMethod() = 0; // setup the method
    virtual void unsetMethod() = 0; // unsetup the method

  // initialize the solution
    virtual void initializeSolution(CFbool isRestart) = 0;

  // compute the space part of residual/jacobian
    virtual void computeSpaceResidual(CFReal factor=1.0) = 0;

  // compute the time dependent part of residual/jacobian
    virtual void computeTimeResidual(CFReal factor=1.0) = 0;

    virtual void applyBC() = 0; // apply boundary conditions
};
```

**Code Listing 2.27: SpaceMethod class definition**

As shown in the sample code above, SpaceMethod inherits from a non-instantiable Method object that provides some configuration functionalities meant to be reused by all its children, namely all the possible BaseMethods in Fig. 2.7.

Figure 2.8 shows a simplified class diagram of a concrete space method ASpaceMethod module. Specific Commands, ASpaceMethodCom, are associated to actions like setup and unsetup (creation and destruction of data needed by the employed scheme), application of boundary conditions, computation of the residual (plus jacobian contributions to the system matrix, in case of implicit schemes) coming from the space and time equation terms, as shown in C.L. 2.28.

All self-registering polymorphic objects, including Commands, are aggregated by SelfRegisterPtrs, i.e. smart pointers with intrusive reference counting that keep ownership on them. In order to take full profit of the self-registration and self-configuration facility which will be described in Sec. 2.5, all ConcreteMethods, including ASpaceMethod, hold the Command names (second entry in the std::pair tuples), which are used as keys for the polymor-
Figure 2.8: OMT diagram of the Method-Command-Strategy pattern applied to a concrete SpaceMethod, called ASpaceMethod.

aphic creation and configuration of the Commands themselves. Let’s consider the constructor of ASpaceMethod:

All the Command names are set to a default that can be overridden by the user in the COOLFluiD input file: these will cause the object requested by the user to be instantiated, if available. As clarifying example, let’s analyze the following fragment of an input file:

```
SpaceMethod = ASpaceMethod
ASpaceMethod.SetupCom = SecondOrderSetup
ASpaceMethod.UnSetupCom = SecondOrderUnSetup
ASpaceMethod.ComputeRHS = JacobRHS
```

This asks to create the SpaceMethod corresponding to the name ASpaceMethod and to select setup and unsetup Commands specific for a second order scheme, which might require the allocation of different or additional data than a first order one. Likewise, the Command with name JacobRHS will be used to compute the residual and jacobian contributions instead of the default one, called StdComputeRHS.

All AComs are parameterized with a policy class [10], ASpaceMethodData,
```cpp
// --- ASpaceMethod.hh --- //
class ASpaceMethod : public SpaceMethod {
  public:
    typedef SelfRegistPtr<Command<ASpaceMethodData>> ACom;
  // constructor, destructor, overridden virtual functions

  private:
  // data to share between ACom commands
  std::pair<ACom, string> m_data;
  std::pair<ACom, string> m_setup;  // setup Command
  std::pair<ACom, string> m_unsetup; // unsetup Command

  // Commands computing the residual/jacobian
  std::pair<ACom, string> m_computeSpaceRHS; // space part
  std::pair<ACom, string> m_computeTimeRHS; // time part

  // Commands that initialize the solution in the domain
  std::vector<ACom> m_inits;
  std::vector<string> m_initsStr; // init Commands names

  // Commands that computes the boundary conditions (bc)
  std::vector<ACom> m_bcs;
  std::vector<string> m_bcsStr; // bc Commands names
};
```

**Code Listing 2.28: ASpaceMethod class definition**

```cpp
// --- ASpaceMethod.cxx --- //
ASpaceMethod::ASpaceMethod(string name) : SpaceMethod(name),
  m_data(new ASpaceMethodData())
{
  m_setup.second = "StdSetup"; // default name
  addConfigOption(SetupCom", "Setup", &m_setup.second);

  m_unsetup.second = "StdUnSetup"; // default name
  addConfigOption(UnSetupCom", "UnSetup", &m_unSetup.second);

  m_computeSpaceRHS.second = "StdComputeRHS"; // default name
  addConfigOption(ComputeRHS", "Compute space residual",
  &m_computeSpaceRHS.second);
  ...
}
```

**Code Listing 2.29: ASpaceMethod constructor**
which groups together all the Strategy objects needed by the Commands to fulfill their job: Strategy1, Strategy2, etc. and some Perspectives, Perspective1, Perspective2, etc. (see 2.3.1), that provide the binding to the physics.

```cpp
// --- ASpaceMethodData.hh --- //
class ASpaceMethodData : public ConfigObject {
    public:
        // constructor, destructor, configuration functions

        // accessor/mutators to Strategies
        SafePtr<Strategy1> getStrategy1() const;
        SafePtr<Strategy2> getStrategy2() const;

        // accessor/mutators to Perspectives
        SafePtr<Perspective1> get Perspective1() const;
        SafePtr<Perspective2> get Perspective2() const;

        // other accessor/mutators ...

    private:

        SelfRegisterPtr<Strategy1> m_strategy1;
        string m_strategy1;
        ...

        SelfRegisterPtr<Perspective1> m_perspective1;
        string m_perspective1;

        // ... the same for all the other objects
};
```

**Code Listing 2.30:** ASpaceMethodData class definition

A more detailed example showing the application of the pattern for implementing a FV method is given in Sec. 5.2.8. Other subclasses of SpaceMethod, such as Finite Element or Residual Distribution, are implemented in a similar way.

### 2.4.1.2 Collaboration between two Methods: ConvergenceMethod and LinearSystemSolver.

We consider now an example of interaction between two different numerical methods. Figure 2.9 shows the collaboration between two abstract methods: ConvergenceMethod, responsible of the iterative procedure, and LinearSystemSolver. In this case, BackwardEuler, an implicit conver-
gence method, delegates polymorphically the solution of the resulting linear system to PetscLSS, which interfaces the PETSc library [73].

**Figure 2.9:** MCS pattern applied to two interacting modules: a Backward Euler convergence method and a Petsc linear system solver.

BackwardEuler makes use of Commands for the setup, unsetup and solution update, while PetscLSS let Commands implement the set up, unsetup and solution of the linear system.

The class definition of ConvergenceMethod is the following:

In the code above, MultiMethodHandle is a lightweight proxy object that hides the knowledge of multiplicity: it controls the access to one or more underlying Methods with the same polymorphic type and dispatches specified actions on all of them sequentially, similarly to what a std::for_each function would do [149]. The purpose of accessing Method objects through MultiMethodHandles is to offer transparent support for weakly coupled simulations, where, in the same process, two or more different linear systems are assembled by one or more SpaceMethods and must be solved one after the other.

SpaceMethod (SM) and LinearSystemSolver (LSS) are the collaborator Methods: a concrete ConvergenceMethod uses them polymorphically via MultiMethodHandles, without knowledge of their concrete type or their ac-
// --- ConvergenceMethod.hh --- //
class ConvergenceMethod : public Method {
public:
    // Constructor, virtual destructor, configuration methods
    virtual void takeStep() = 0; // Take one timestep
    virtual void setMethod() = 0; // Sets up private data
    virtual void unsetMethod() = 0; // Clears up private data

    // Sets collaborator methods (SpaceMethod, LinearSystemSolver)
    void setCollaborator(MultiMethodHandle<SpaceMethod> spaceMtd);
    void setCollaborator(MultiMethodHandle<LinearSystemSolver> lss);

protected: // data
    // Space Method used to compute spatial discretization
    MultiMethodHandle<SpaceMethod> _spaceMtd;

    // LinearSystemSolver used to solve the linear system (if any)
    MultiMethodHandle<LinearSystemSolver> _lss;

    // handle to the global storage of states
    DataHandle<GLOBAL,CFreal> _statedata;

    // handle to the global storage of nodes
    DataHandle<GLOBAL,CFreal> _nodedata;
};

**Code Listing 2.31:** ConvergenceMethod class definition

tual number.
As a result, concrete collaborator **Methods** are completely interchangeable with other ones with the same polymorphic type. When running in parallel, in the function **synchronAndComputeRes()** the global storage of state vectors and nodal coordinates are accessed via **DataHandle** and asked to synchronize the underlying parallel array, as shown in C.L 2.32.

In a parallel simulation, the norm of the residual is calculated with a collective operation, between the beginning and the end of the synchronization process, in order to overlap communication and computation and maximize the efficiency of the operation. This is one of the only few COOLFluiD code fragments that makes direct use of the global **DataHandles** and needs to conditionally apply a different treatment in parallel or in serial. We consider
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```cpp
// --- ConvergenceMethod.cxx --- //
void ConvergenceMethod::synchAndComputeRes(CFbool computeResidual)
{
    // after each update phase states and nodes
    // have to be synchronized
    if (isParallel()) {
        m_statedata->beginSync();
        m_nodedata->beginSync();
    }
    // computation of the residual
    if (computeResidual) {...}

    if (isParallel()) {
        m_statedata->endSync();
        m_nodedata->endSync();
    }
}
```

**Code Listing 2.32**: ConvergenceMethod synchronization

Now a backward Euler time stepper, implemented as a subclass `BwdEuler` of the `ConvergenceMethod`:

```cpp
// --- BwdEuler.hh --- //
class BwdEuler : public ConvergenceMethod {
public:
    // constructor, virtual destructor, configuration methods
typed SelfRegistPtr<Command<BwdEulerData> > BwdEulerCom;

    void takeStep(); // Take one timestep
    void setMethod(); // Sets up private data
    void unsetMethod(); // Clears up private data

private:
    // data shared by BwdEuler Commands
    std::auto_ptr<BwdEulerData> m_data;
    std::pair<BwdEulerCom,string> m_setup; // set up
    std::pair<BwdEulerCom,string> m_unSetup; // unsetup
    std::pair<BwdEulerCom,string> m_updateSol; // update solution
};
```

**Code Listing 2.33**: BwdEuler class definition

A different `Command` and its name are associated to each one of the pure virtual functions declared by the parent `ConvergenceMethod`. We report here after the implementation of `takeStep()`, where the polymorphic usage of the collaborators (SM and LSS) and of the solution updating `Command` is
shown.

```cpp
// --- BwdEuler.cxx --- //
void BwdEuler::takeStep()
{
    // compute spatial and Jacobian contributions for
    // the spatial and time dependent term of the equations
    m_spaceMtd.apply(mem_fun(&SpaceMethod::computeSpaceResidual));
    m_spaceMtd.apply(mem_fun(&SpaceMethod::computeTimeResidual));

    // solve the resulting linear system
    m_lss.apply(mem_fun(&LinearSystemSolver::solveSys));

    m_updateSol.first->execute(); // update the solution

    // synchronize nodes and states, compute the intermediate residual
    ConvergenceMethod::synchAndComputeRes(true);
}
```

**Code Listing 2.34:** BwdEuler takeStep() function

The code is readable, concise, independent from the actual type and number of SM or LSS, extremely flexible, since each part of the algorithm (Commands or Methods) can be replaced at run-time without any performance overhead. In fact, the frequency of virtual calls in question is exceptionally low and, therefore, does not have an impact on the run-time speed.

The class definitions of the parent LSS and of its subclass PetscLSS are shown in C.L. 2.35 and 2.36.

PetscLSS delegates tasks to specific Commands, PetscLSSCom, sharing some data (PetscLSSData), such as references to the (parallel) Petsc matrix and the (parallel) Petsc vectors involved in the solution of the linear system. Only the PetscLSSCom can make direct use of PETSc [73] objects like PC or KSP, which are aggregated by PetscLSSData. The knowledge of a specific LSS, PetscLSS, in this case, is not assumed anywhere in the numerical modules, thanks to the use of abstractions such as LSSMatrix, BlockAccumulator and LSSIdxMapping. LSSMatrix is the parent system matrix from which PetscMatrix derives. BlockAccumulator bundles blocks of values to be inserted in the matrix. LSSIdxMapping stores a mapping from the local numbering to an optimal LSS-dependent global one.

As represented in Fig. 2.9, all the involved Commands have acquaintance of MeshData, which provides access to DataStorage and DataHandles, and can therefore use and modify bulk data, qualified by name and type, as explained in 2.2.1. In other words, while, on the one hand, each ConcreteMethodData,
such as PetscLSSData and BackwardEulerData, allows intra-Method sharing of ConcreteMethod-dependent data among Commands, on the other hand, MeshData is the vehicle for inter-Method data exchange.

The sample code in C.L. 2.38 should help clarifying the last statement in the case of two Commands, namely UpdateSol in BackwardEuler and SolveSys in PetscLSS.

Data stored in MeshData are allowed to cross the Method scope and can be used in Commands belonging to different Methods. The keys for the data exchange are the storage name and type, that must both match in all the method Commands that need the same data.

In parallel simulations, nothing changes, since the Commands implementing numerical algorithms work only with LOCAL data, as they would do in a serial run. Functions that demands access to the GLOBAL storage and perform some parallel action, like the above mentioned synchAndComputeRes(), are exceptional.

Furthermore, the example of the collaboration between ConvergenceMethod and LSS demonstrates the suitability of the MCS pattern to interface existing libraries, PETSc in this case, without exposing any detail of their actual
implementation to their clients.

The Trilinos package [72] has also been successfully integrated in COOLFluiD by means of the MCS pattern. While the class definitions of the corresponding concrete linear system solver Method and the related Commands are basi-
cally similar to the Petsc’s ones, the interface of TrilinosLSSData is defined in C.L. 2.40.

In summary, the application of the MCS pattern helps to encapsulate both the single numerical algorithms and their collaborations with other classes of schemes. This contributes to enforce the multi-component-oriented character of the COOLFluiD framework, where each component can be replaced by another one with the same polymorphic type, without affecting the client code.
2.5 Dynamical plug-ins

In COOLFluiD, each concrete numerical method or physical model is enclosed in a separate *plug-in* library or module. This plug-in policy, which provides our platform with significant modularity and extensibility, relies heavily on two complementary techniques, namely *self-registration* and *self-configuration* of objects, whose basic principles re explained in this section.

### 2.5.1 Self-registering objects.

The self registration of objects, pioneered by [19] in a C++ context, automatizes the creation of polymorphic objects so that

1. the details of the concrete types are not exposed to the interface of the core code;
2. all possible objects are created using the same interface;
3. the core implementation should be unaffected by future unpredictable extensions of the code.

As a result, both implementation and compilation dependencies are considerably reduced, allowing developers to enclose each new functionality as a separate component, compile it separately and loading it in dynamically on demand into the platform, without even needing a partial recompilation. Our implementation of this technique, which is of great help in easing the
integrability of new components in COOLFluID, has been inspired by [39],
but we have evolved it into a more sophisticated and flexible approach, only
partially explained in [74, 78] which is described here after.
A singleton [48] Factory class for a generic polymorphic object type (here
represented by the template parameter OBJ) is defined in C.L. 2.41.

```cpp
// --- Factory.hh --- //

template <class OBJ>
class Factory {
  public:
    // get a reference to the Factory itself
    static Factory<OBJ>& getInstance()
      { static Environment::Factory<BASE> obj; return obj; }
  
    // register a provider
    void regist(Environment::Provider<OBJ>* provider)
      { m_map[provider->getName()] = provider; }

    // get the provider corresponding to the given key name
    SafePtr<typename OBJ::PROVIDER> getProvider(string pName)
      { return dynamic_cast<typename OBJ::PROVIDER*>(
          m_map.find(pName)->second); }

  private:
    std::map<string, Provider<OBJ>*> m_map; // providers database
};
```

**Code Listing 2.41**: Definition of Factory class

Factory encapsulates an associative container, a std::map, which stores pairs
key-value, where the key is a literal string and the value is a pointer to a
Provider for objects of polymorphic type OBJ. Each Provider of a poly-
morphic object of dynamic type OBJ registers its name and itself in the
corresponding Factory class at construction time, as shown in C.L. 2.42.
Another intermediate abstract class, called ConcreteProvider, needs to be
defined: it derives from Provider and defines a virtual create function. In
fact, several template specializations [160] of ConcreteProvider are pro-
vided, one for each foreseen number of parameters (0,1,2) accepted by the
create function. Two of those class specializations are shown in C.L. 2.43.
The type SelfRegistPtr appearing in C.L. 2.43 is just a proxy pointer class
for handling self-registering objects in safety.
The implementation for the create functions defined by ConcreteProvider
is given by ObjectProvider in C.L. 2.44, which also needs to be partially
specialized for accomplishing the task.
In particular, `ObjectProvider` is parameterized with three template arguments: the concrete type `CONCRETE`, its corresponding polymorphic type `OBJ` and the number of arguments for the creational function.

We now consider an example to show how the whole machinery works in practice. We define an abstract class `Shape` whose constructor accepts one string parameter. In order to be able to apply the self-registration of all its possible derived classes, the typedefs `PROVIDER` and `ARG1` must be explicitly
// --- ObjectProvider.hh --- //

template <class CONCRETE, class OBJ, int NBARGS = 0>
class ObjectProvider : public OBJ::PROVIDER {
public:
  // implementation of the create function with 0 arguments
  SelfRegStmt<OBJ> create()
  { return SelfRegStmt<OBJ>(new CONCRETE(), this); }
};

template <class CONCRETE, class OBJ>
class ObjectProvider<CONCRETE, OBJ, 1> : public OBJ::PROVIDER {
public:
  // implementation of the create function with 1 arguments
  SelfRegStmt<OBJ> create(typename OBJ::ARG1 arg)
  { return SelfRegStmt<OBJ>(new CONCRETE(arg), this); }
};

Code Listing 2.44: ObjectProvider class

provided with public access in the Shape class definition, as shown in C.L. 2.45.

// --- Shape.hh --- //

class Shape {
public:
  typedef ConcreteProvider<Shape, 1> PROVIDER;
  typedef string& ARG1;
  Shape(string& name);
  virtual double getVolume() = 0;
};

Code Listing 2.45: Shape class

At this point, whatever shape object (Cube, Circle, Polyhedron, etc.) deriving from the parent Shape can be self-registered by instantiating a global variable with type equal to the corresponding ObjectProvider, somewhere in the code (e.g. in the implementation file):
The registration of all available ObjectProviders occurs before entering the main() function of the code, since Factories live on the static memory and ObjectProviders are instantiated as global variables.
In our example, the string "Cube" (or "Circle"), accepted by the corresponding provider constructor, can then be used as a key to ask the singleton Factory to query its database and create the corresponding polymorphic
// --- Cube.hxx --- //
2 ObjectProvider<Shape, Cube, 1> cubeProvider("Cube");

// --- Circle.hxx --- //
5 ObjectProvider<Shape, Circle, 1> circleProvider("Circle");

**Code Listing 2.46:** Instatiation of a ObjectProvider for a Cube in the implementation file

object whenever needed.

1 string keyName = shapeName(); // can be "Cube" or "Circle" etc.
4 SelfRegister<Shape> shape = Factory<Shape>::getInstance().
6 getProvider(keyName)->create(keyName);

**Code Listing 2.47:** Creation of a polymorphic shape object

This example shows that whatever the actual shape, the way to create it remains unique and the piece of code in C.L. 2.47 doesn’t need to be changed anymore even if new shapes are developed and integrated in the platform. In COOLFluiD, all polymorphic objects are created with this technique, once their corresponding names (e.g. `keyName` in the example in C.L. 2.47) are read from the CFcase input file.

### 2.5.2 Self-configurable objects.

In COOLFluiD, objects can be self-configurable [78, 80, 131], i.e. they can create and set their own data. In order to make an object self configurable, some steps have to be followed. First, the object must derive from the abstract parent class `ConfigObject`, which implements the configuration algorithm [131].

Second, a static member function `defineConfigOptions()` must be defined and implemented. The implementation consists in declaring the type, name and description for each one of the options, as in the illustrative example in C.L. 2.48 for a `TimeStepper` object:

Once declared, the options must be linked to the actual data to configure and this is done inside the constructor, as in the example in C.L. 2.49.

In particular, the strings "CFL" and "LSS" in our example are the configuration keys and they are used to map the values of the private data `m_cfl` and `m_lssName` with the ones read from the configuration case file (CFcase).
// --- TimeStepper.cxx --- //
void TimeStepper::defineConfigOptions(OptionList& op)
{
    op.addConfigOption<
double>
        ("CFL", "CFL number");
    op.addConfigOption<string>
        ("LSS", "Linear System Solver name");
}

Code Listing 2.48: Example of implementation of addConfigOption() function for a self-configurable TimeStepper

// --- TimeStepper.cxx --- //
TimeStepper::TimeStepper() : ConfigObject()
{
    addConfigOptionsTo(this);
    m_cfl = 1.0; // default CFL value
    setParameter("CFL", &m_cfl);
    m_lssName = "PETSC"; // default LSS name
    setParameter("LSS", &m_lssName);
}

Code Listing 2.49: Example of configuration directives in the constructor of a self-configurable TimeStepper

In order to configure our TimeStepper object, the end-user should write something like

TimeStepper = BwdEuler  # name of the concrete TimeStepper
BwdEuler.CFL = 3.5     # user-defined CFL number
BwdEuler.LSS = TRILINOS # user-defined linear system solver

in the CFile case file. BwdEuler is the self-configuration value for TimeStepper which is the configuration key for the homonymous polymorphic object. BwdEuler is also the self-registration key (see Sec. 2.5.1) for the concrete time stepper class that will then be instantiated and will configure itself with the specified CFL number and linear system solver name. In COOLFluiD, Methods, Commands, Strategies, PhysicalModels etc. are all self-configurable objects and this, together with the self-registration technique, gives the end-user full power on their settings. The actual implementation of the described technique, due to [131], allows users to input whatever
kind of data (including analytical functions) from file, environmental variables or command line options, but also to define interactive parameters or to substitute on-the-fly polymorphic objects during the simulation.