A PURE VIRTUAL APPROACH FOR MANAGING PLATFORM PORTABILITY ON HYBRID SUPERCOMPUTERS

Xavier ÁLVAREZ-FARRÉ¹, Andrey GOROBETS², Àdel ALSATI¹ AND F. Xavier TRIAS¹

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¹ Heat and Mass Transfer Technological Center, Technical University of Catalonia (UPC)
² Keldysh Institute of Applied Mathematics, Russian Academy of Science (RAS)
Background
Research at CTTC Laboratory

The Heat and Mass Transfer Technological Center (CTTC) in Technical University of Catalonia (UPC) has been working on CFD for more than 20 years. Our research activities are focused on two main lines:

- Fundamental research on fluid dynamics and heat and mass transfer phenomena.
- Applied research on thermal and fluid dynamic optimization of thermal system and equipment.
HPC at CTTC Laboratory

Emergent Technologies


Single-core CPU Multi-core CPU

GPU

MIC

CTTC resources and software

DPC
Sequential structured

STG
Parallel structured

Deep source of applied and fundamental research

Scalability up to 100k+ cores

Ooops! 404

![Image 1](image1)
![Image 2](image2)
![Image 3](image3)
![Image 4](image4)
Emerging architectures

*Continuous enhancement*

in hardware technologies enables scientific computing to advance incessantly to reach further aims. After hitting petascale speeds in 2008, several organisations and institutions began the well-known global race for exascale high-performance computing (HPC).
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Massively-parallel devices of various architectures are being incorporated into the newest supercomputers, causing the hybridisation of HPC systems and making the design of computing applications a rather complex problem. The kernels conforming the algorithms must be compatible with distributed- and shared-memory SIMD and MIMD parallelism, and with stream processing (SP), which is a very restrictive parallel paradigm.
Is it necessary to use the new hardware architectures?

• In our opinion, yes. New hardware is designed to overcome the power constraint in the context of the exascale challenge.
The future of scientific computing codes

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¿Do we need to change the way we look at scientific computing in general?

- In our opinion, yes. There is a huge amount of hardware architectures and it is difficult to determine which are going to prevail. The scientific computing software should be designed somehow so that computing kernels can operate independently.
The algebra-based approach

Stencil-based

Traditionally, the stencil-based implementations are used by the scientific computing community. These implementations arise straightforward from the formulation of the numerical method. However, they require **specific stencil sweeps and data structures** for each numerical method.

Algebra-based

Algebra-based implementations only rely on a reduced number of **universal algebraic kernels and data structures**, allowing the use of standard optimised libraries and, therefore, providing portability. As a counterpart, the formulation of the numerical method becomes more complex and could even lead to an increase in the number of operations.
The algebra-based approach

Following an algebra-based approach, we replace the traditional stencil data structures and sweeps by algebraic data structures and kernels\(^1\). For instance, the algebra-based, finite-volume discretisation of NS and continuity equations on an arbitrary collocated mesh can be written as\(^2\)

\[
\nabla \cdot \mathbf{u} = 0, \quad \partial_t \mathbf{u} + (\mathbf{u} \cdot \nabla)\mathbf{u} - \frac{1}{Re} \Delta \mathbf{u} + \nabla p = 0,
\]

\[
M\mathbf{u}_s = 0_c, \quad \Omega d_t \mathbf{u}_c + M\mathbf{U}_s\mathbf{u}_c + \mathbf{D}\mathbf{u}_c - M^\top p_c = 0_c.
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1. The discrete variables are stored in vectors and the discrete operators in sparse matrices.
2. The numerical method results fully integrated into the data structures somehow so that computing kernels can operate independently.
3. The discrete operators can be built directly from the inherent incidence matrices that define the mesh mimicking the properties of the continuum operators.

The HPC\(^2\) framework
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The HPC$^2$ (Heterogeneous Portable Code for HPC) is a fully-portable, algebra-based framework with many potential applications in the fields of computational physics and mathematics. Its algebraic approach combined with a multilevel MPI + OpenMP + OpenCL + CUDA parallelisation naturally provides modularity and portability.
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Using our algebra-based framework, a numerical algorithm relies on a few computing kernels, grouped in the following three types of algebraic operations:

- **spmv**: sparse matrix-vector operations.
- **nary**: n-ary vector operations require n input vectors and return an output vector.
- **sred**: reduction operations require n input vectors and return a scalar value.

Besides, having a reduced set of simple kernels, the arithmetic intensity of an algorithm⁴ can be estimated easily and thus its relative performance.

<table>
<thead>
<tr>
<th>Kernel</th>
<th>Work</th>
<th>Read</th>
<th>Write</th>
<th>AI</th>
</tr>
</thead>
<tbody>
<tr>
<td>spmv</td>
<td>13N</td>
<td>64N</td>
<td>8N</td>
<td>13/72</td>
</tr>
<tr>
<td>axpy</td>
<td>N</td>
<td>8N</td>
<td>8N</td>
<td>1/16</td>
</tr>
<tr>
<td>sdot</td>
<td>2N</td>
<td>16N</td>
<td>8</td>
<td>1/8</td>
</tr>
</tbody>
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The HPC$^2$ pure-virtual core

The HPC$^2$ is organised as follows: \texttt{vector}, \texttt{matrix} and \texttt{unit} are pure virtual classes; derived classes may be developed for OpenMP, OpenCL, CUDA, etc. The handlers \texttt{VECTOR}, \texttt{MATRIX} and \texttt{NODE} may contain several instances of virtual objects.
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The first-level domain decomposition distributes the workload among the computing nodes. Subdomain elements are classified into *inner* and *interface* categories. Consequently, the adjacent elements from other subdomains form a *halo*.
The second-level domain decomposition distributes the workload of each MPI process among its computing hardware, namely *host* and *devices*. The interface and halo elements are further classified as *external* and *internal*. 

Second-level decomposition, intra-node parallelisation

- **Subdomain 1**
  - Host
  - D1
  - D2
  - Inner
  - Exterior Halo
  - Exterior Interface
  - Interior Interface
  - Interior Halo

- **Subdomain 2**
  - Host
  - D1
  - D2
  - Inner
  - Exterior Halo
  - Exterior Interface
  - Interior Interface
  - Interior Halo

Second-level decomposition reduces the volume of the expensive device-host-MPI-host-device communications several times!
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Multithreaded overlap strategies

The strategies for an efficient heterogeneous execution of large-scale simulations on hybrid supercomputers that are part of the HPC$^2$ core are shown in the flowchart below:
Performance analysis
The roofline model is an intuitive performance analysis model used to estimate the maximum performance of a given computing kernel depending on the actual hardware specifications: the memory bandwidth, \( \pi \), and the peak performance, \( \beta \):

\[
R_k = \min(\pi, \beta \cdot AI_k).
\]
The heterogeneous performance study shows an increase of 32% compared to GPU-only mode, which corresponds to a **98% of heterogeneous efficiency** if compared to the sum of the performance of the CPU-only and the GPU-only modes.
Multi-node, heterogeneous performance study

The **strong scalability** study (ran on nodes equipped with an Intel E5-2697v3 and a NVIDIA Tesla K40M) shows that the simple overlap strategy improves the performance by hiding the communications. The scalability decays faster in the heterogeneous mode because the computational load per GPU is smaller and the CPU is loaded with computations which may interfere with MPI library routines, reducing the overlapping operational range.
Conclusions
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- On the one hand, this allows researchers to develop different code blocks avoiding interferences, reduces the generation and propagation of errors and facilitates debugging. The progress of a modular code may be accumulative.
- On the other hand, modular applications are user-friendly and more comfortable for porting to new architectures (the fewer the kernels of an application, the easier it is to provide portability). Furthermore, if the majority of kernels represent linear algebra operations, then standard optimised libraries (e.g. ATLAS, clBLAST) or specific in-house implementations can be used and easily switched.
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- It also requires a **larger use of data** since the numerical method is completely integrated into data structures.
- The implementation of complex boundary conditions is challenging. Are the boundary conditions an operator property or variable one?
Thank you for your attention