Introduction

Heterogeneity is becoming a fact of life in HPC, largely driven by demands for increased parallelism and power efficiency over what traditional CPUs can provide. However, extracting the full performance of heterogeneous systems is non-trivial and requires architecture expertise. Retrofitting existing codes for heterogeneity is tedious and error-prone, architecture experts are in short supply, and accelerators are moving targets.

Therefore, a single API for transparently executing optimized code on accelerators with minimal intervention is needed for scientific productivity.

Design

"Make-your-own" library from modular building blocks

Include only needed plugins and backends

User Apps

Purpose-built to provide single API to accelerated kernels for current and future devices

C API

Timer Plugin

MPI Plugin

Fortran 2003 API

Runtime control over backends, plugins, and accelerator device selection

CUDA Backend

OpenCL Backend

Future Device Backends

Backend Layer

Plugins

Automatic Profiling

If compiled in, all kernels and transfers are timed behind the scenes automatically

Environment variable controls verbosity

Fortran Compatibility

Verboso pass-by-reference C API, usable with ISO_C_BINDINGs

Fortran 2003 wrapper to verbose API provides simplified unified calling convention for supported real and integer types

Uses C API internally, so all runtime control variables work equivalently

On-device packing of ghost regions

GPU Direct option, if available with MPI backend

Fallback to host-staged transfers otherwise

Transparently exchange device buffers between processes, regardless of backend

Performance

All tests performed in a single system containing 2x Intel Xeon W5500 Quad-core CPUs, 2008 RHEL, and Tesla 2050 GPU

Future Work

Continue expanding the API's provided set of kernels and backends with other primitive operations underlying fluid simulations, i.e. Krylov solvers, stencil computations, and various preconditioners

Generalize operations to work on non-3D data, and add primitives for computations on unstructured grids

Generate a third automatically runtime-scheduled backend to transparently execute code across entire node, or a la CoreTSAR[3].

Related Efforts

Solver Frameworks

OpenFOAM [1] Pros: Support for useful pre- and post-processing (mesh generation and visualization), many solvers for many domains

Cons: No internal accelerator support; framework-centric development; cumbersome API and "case" construction

PARALUTION [2] Pros: Many matrix storage formats; many solvers; many preconditioners; support for OpenMP, CUDA, and OpenCL on CPUs/GPUs and MC, plugins for Fortan and OpenFOAM

Cons: Framework-centric development; interoperating with existing code base; no MPI support (yet); asynchronous operations only on CUDA; lack of non-destructive copy-to/from C arrays

Solver Libraries

MAGMA [3] Pros: Full BLAS and LAPACK support for CUDA, OpenCL, and MIC; support for several factorizations and eigenvalue problems; smart scheduling of hybrid CPU/GPU algorithms with QMRR directed acyclic graph scheduler; Multi-GPU methods

Cons: CUDA, OpenCL, and MIC variants are separate implementations; no internal MPI support; MKL/ACML dependency poorly documented and cumbersome

Trilinos [4] Pros: Massive set of capabilities beyond linear algebra, solvers, and meshes; built-in distributed memory support; some preliminary CUDA/MIC work (e.g. Kokkos, Phalanx, Tetra packages)

Cons: Redundancies of capability between packages; breadth of packages difficult to navigate for newcomers

Accelerated Backends

The heavy-lifters of the library, selected at runtime by a "mode" environment variable from those included at compile-time

Include implementations of all C API-supported kernels for a single accelerator model

Standalone libraries in their own right can be used and distributed separately from the top-level API, as long as they are API compliant, supporting community development of closed- or open-source alternatives

CUDA and OpenCL, providing access to the most popular accelerators

Currently support simple operations on subsets of 3D dense matrices: reduction-sum, dot-product, 2D transpose, pack/unpack of subregions

More kernels from computational fluid dynamics in the pipeline; extensions to other domains to follow, simply a matter of adding necessary kernels

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