Application-Driven Requirements for Node Resource Management in Next-Generation Systems

Edgar A León Balazs Gerofi Julien Jaeger Guillaume Mercier Rolf Riesen Masamichi Takagi Brice Goglin



13 November 2020











LLNL-PRES-816236

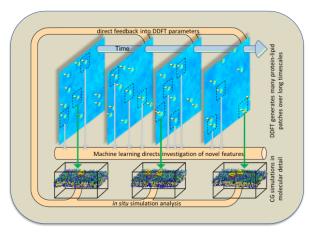
LLNL-PRES-816236

The Committee of the U.S. Department of Energy by Lawrence Livermore National Laboratory under contract DE-AC\$2-07N-927344

Lawrence Livermore National Security. LLC

Increasingly complex workflows are pushing the limits of HPC software environments

- Multi-physics simulations
 - National security applications
- Data science workloads
 - TensorFlow
 - PyTorch
 - LBANN
- Multi-kernel applications
 - Weather prediction models
- Cognitive simulations
 - Conventional HPC + Deep-Learning

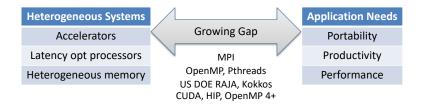


A Massively Parallel Infrastructure for Adaptive Multiscale Simulations.

Di Natale et al., SC 2019

Heterogeneous supercomputing systems pose challenges to application and library developers

- Users must utilize a combination of programming models and runtimes
- Components assume dedicated resources
 - Coordination is left to application and library developers

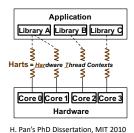


Poor or no coordination of resources among node-local components



While not new, this problem is exacerbated by workflow complexity and system complexity

- Promising research in this area
 - Lithe, QUO
 - Argo, Hobbes, multi-kernels
 - Resource and workflow managers



Yet, problem is getting worse

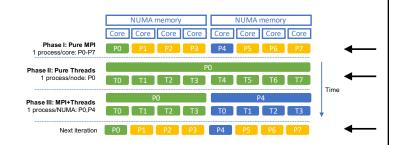
Our work:

- Understand requirements of current and emerging applications
- Group challenges into a handful of themes we can study
- Propose a strawman solution for application composition

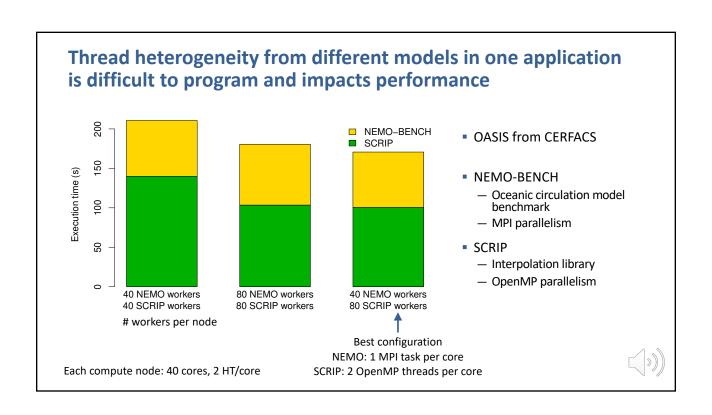


Climate modeling applications integrate multiple kernels with different runtime configurations

- Multiple domains and models
 - Ocean physics
 - Atmospheric physics
 - Biogeochemistry
- Model-specific kernels are developed independently
 - Need to exchange data
 - May use different runtimes
 - May have different optimal points

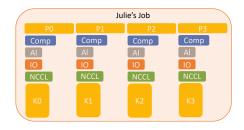






Coordinating multiple components, each with a different type and number of workers, is challenging and error-prone

- Inertial Confinement Fusion
 - LLNL's National Ignition Facility
 - Livermore Big Artificial Neural Network
- I/O
 - C++ threads on CPU
- Compute
 - OpenMP threads on CPU
 - GPU kernels
- Communication
 - Aluminum Pthreads
 - NCCL Pthreads



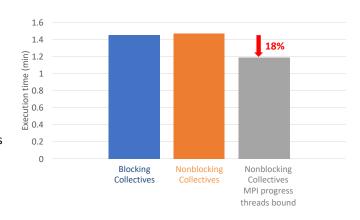






Placement of compute and utility threads is an important factor to enable computation and communication overlap

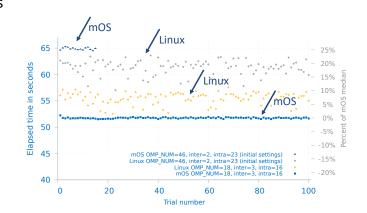
- Water flow simulations
 - GeoFEM from University of Tokyo
- Overlap comm & computation
 - Nonblocking MPI collectives
 - MPI's progress threads
- Considerations
 - Placement of utility threads matters
 - MPI library has no visibility into workers from other libraries



8 MPI tasks/node, 8 OpenMP threads/task 68 cores per compute node

Conflicting directives from different parts of the software stack may hinder performance and the ability to optimize

- Deep learning in Cancer problems
 - CANDLE benchmark
 Pilot 3 data set from ECP
 - TensorFlow + Intel MKL-DNN
- TensorFlow
 - Two thread pools to stage work
 - User controls size but not placement
- Intel MKL-DNN
 - OpenMP threads
 - User controls size and placement



Compute node with 48 cores

Our work:

- Understand requirements of current and emerging applications
- Group challenges into a handful of themes we can study
- Propose a strawman solution for application composition

We created 4 themes to capture application requirements and simplify their study

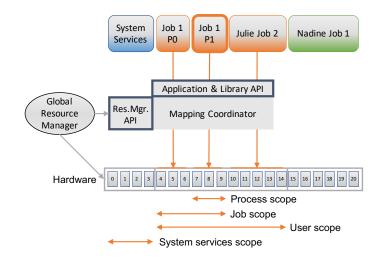
	CANDLE	GeoFEM	NWChem	NEMO BENCH	Hydro	ICF LBANN	SCALE- LETKF
Multiple types of workers							
Dynamic compute workers							
Dynamic utility workers							
Remapping of tasks/threads							
Multiple applications							
MPI							
OpenMP							
POSIX threads							
NVIDIA CUDA							
C++ threads							



Our work:

- Understand requirements of current and emerging applications
- Group challenges into a handful of themes we can study
- Propose a strawman solution for application composition

Our *Mapping Coordinator* can provide the functionality needed to meet application demands

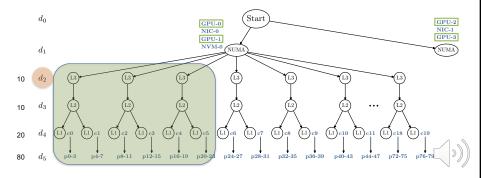


- Provides two-prong interface
 - Low-level / high-level functions
- Reconfigures worker types dynamically
- Manages node-local HW
 - Affinity and binding
 - Query availability
 - Request & release
 - Arbitrate access
- Coordinates with the global resource manager

Scopes and mapping policies are high-level abstractions to help application developers

- Mapping policies
 - High-level interface
 No HW topology specific info
 Portable and applied dynamically
 - Low-level interface

- mpibind
 - Simple interface: # workers
 - Memory driven
 - Heterogeneous architectures
 - Hybrid applications



Enabling application composition will be paramount for emerging science applications on tomorrow's systems

- Components assume dedicated resources
 - Coordination left to app/library developers
- Problem not new, but is getting worse
 - Application workflows are more complex
 - Systems are more heterogeneous

- Identified challenges based on real applications
 - Multiple, uncoordinated types of threads
 - Dynamic work by auxiliary libraries
 - Rebalance and remap of workers
 - Multiple applications working together
- Proposed a Mapping Coordinator strawman to mitigate challenges





