



Performance Portable Assembly For Plasma Fluid Equations

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ECCOMAS 2016

Tuesday June 7th, 2016



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Problem Description

Prediction of plasma processes at moderate to high densities

- Multifluid plasma equations – Electrons, ions and neutrals
 - Six-dimensional Boltzmann equations too high of dimensionality
 - Particle methods too expensive due to plasma frequency
 - Sets of 5 or 13 coupled PDEs with stiff source terms need to be solved
- Equations need to be composable and maintainable

Analysis Beyond Forward Simulation

- Forward solves are not enough – we want to explore complex solution spaces:
 - Simultaneous analysis and design adds requirements (typically sensitivities)
 - Do not burden analysts/physics experts with analysis algorithm requirements: i.e. programming sensitivities for implicit solvers, optimization, stability, bifurcation analysis and UQ

Engine must be flexible, extensible, maintainable and EFFICIENT!



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Directed Acyclic Graph-based Assembly


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Template-based Generic Programming

Engine must be flexible, extensible, maintainable and EFFICIENT!



Drekar/Panzer: Algorithms and Software

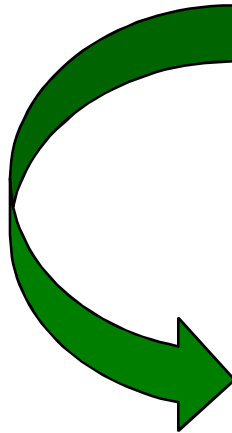
Drekar: Algorithms research application

- Targets coupled multi-physics
- Large scale simulation (>100k cores)
- Advanced algorithm demonstration

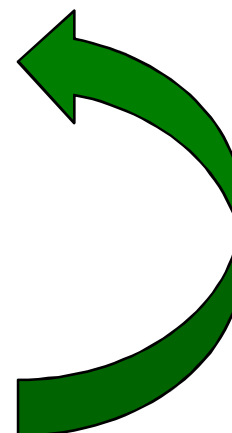
A co-dependent development relationship

Panzer: Multi-physics assembly engine

- Finite element focused (currently)
- Embedded analysis (AD, Sensitivities)
- Technology sharing and deployment



Drekar drives Panzer
requirements and
design goals



Panzer provides
Drekar flexible
infrastructure and core
technologies

John Shadid
Roger Pawlowski
Eric Cyr
Edward Phillips
Tim Wildey

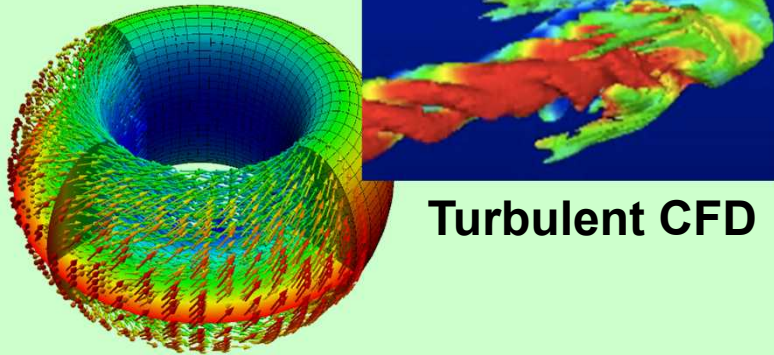
Tom Smith
Paul Lin
David Sondak
Paula Weber
Richard Kramer

Allen Robinson
Matt Bettencourt
Sidafa Conde
Ben Seefeldt
Chris Siefert

Andrew Bradley
Greg von Winckel
David Hensinger

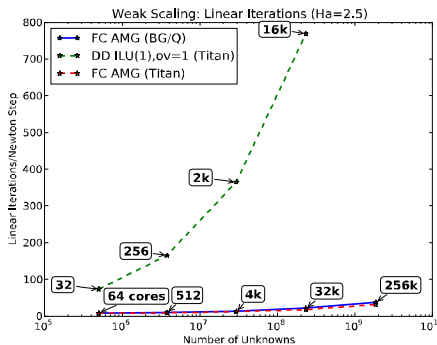
Drekar/Panzer: Capabilities

Applications



Turbulent CFD

Magnetohydrodynamics



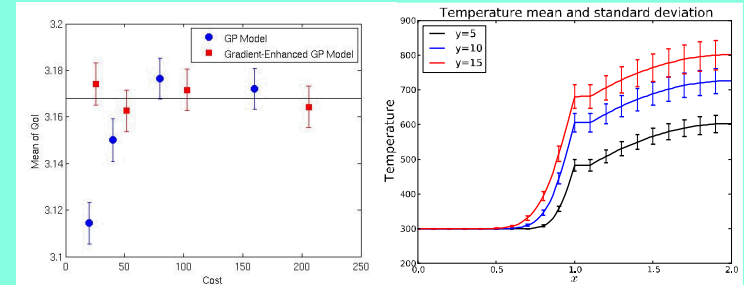
**Algebraic Multigrid
(>100k cores)**

$$\mathcal{A} = \begin{bmatrix} I & \\ BF^{-1} & I \end{bmatrix} \begin{bmatrix} F & B^T \\ S & \end{bmatrix}$$

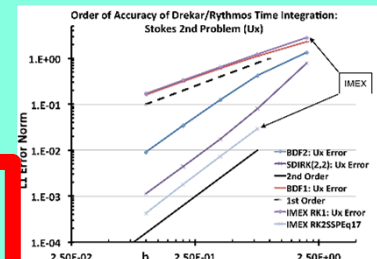
$$S = C - BF^{-1}B^T$$

**Block
Preconditioning**

Discretizations & Algorithms

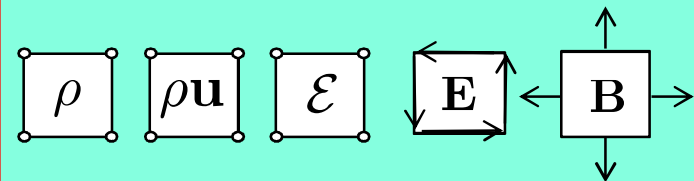


Uncertainty Quantification



IMEX

**PDE Constrained
Optimization**

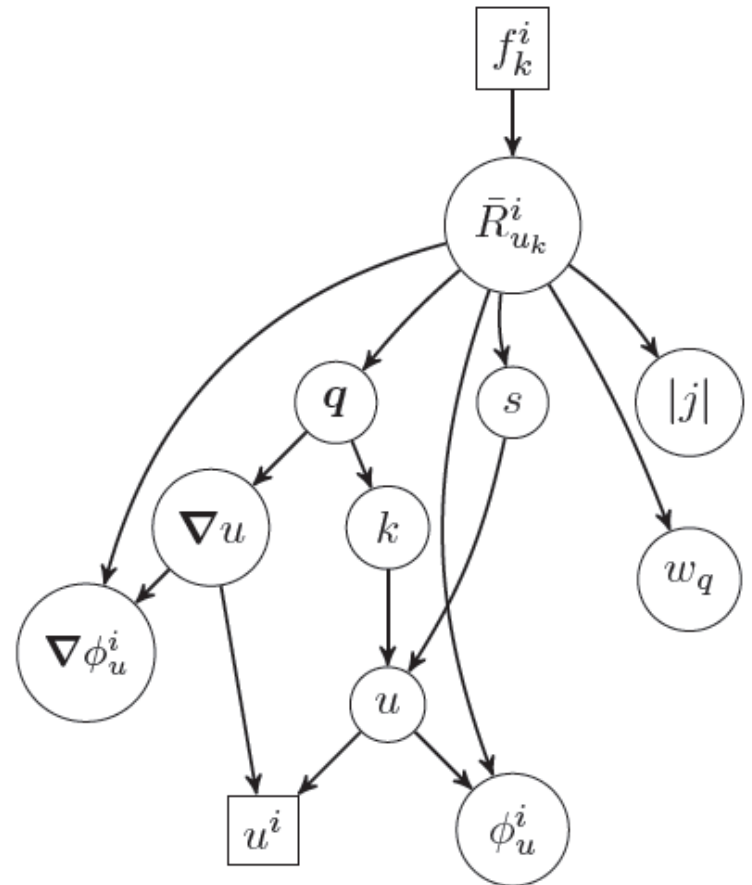


Compatible Discretizations

Li
E

- Decompose a complex model into a graph of simple kernels (functors)
 - Decomposition is NOT unique
- Supports rapid development, separation of concerns and extensibility.
- A node in the graph evaluates one or more **fields**:
 - Declare fields to evaluate
 - Declare dependent fields
 - Function to perform evaluation
- Separation of data (Fields) and kernels (Expressions) that operate on the data
 - Fields are accessed via multidimensional array interface (shards or kokkos)

$$R_u^i = \int_{\Omega} [\phi_u^i \dot{u} - \nabla \phi_u^i \cdot \mathbf{q} + \phi_u^i s] \, \mathrm{d}\Omega$$



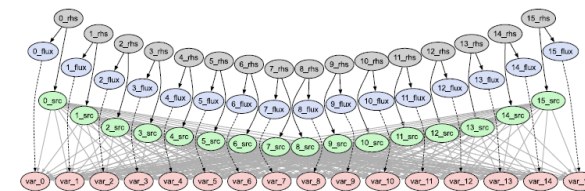
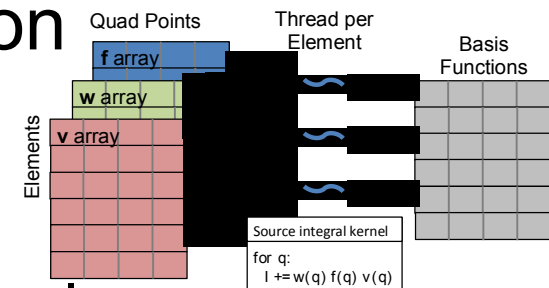
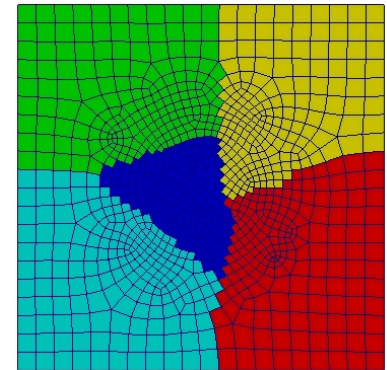


DAG Evaluation

- The problem domain is broken into worksets
 - Smaller number of elements which fit into cache or other memory structures
 - The entire DAG is evaluated for a single workset
 - Each math kernel can be threaded over the workset elements
- Worksets are filled with a gather operation
- The workset's contributions are accrued in a matrix with a scatter operation
- All intermediate calculations are independent of if they are computing a Jacobian or residual or sensitivity

Parallelization Strategies

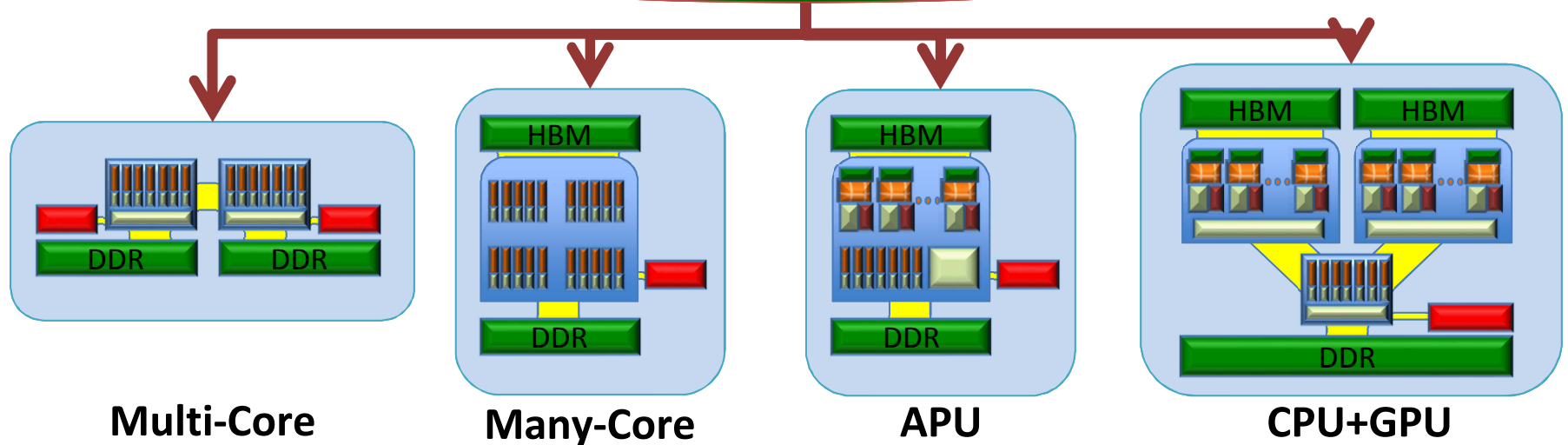
- MPI – Domain decomposition
 - This is the traditional approach
 - Not covered in this talk
- Threading – On node parallelization
 - More scalable for local cores
 - Extensible to GPUs/Cuda
 - Threads team up to accelerate a kernel
- Kernel parallelization
 - Different kernels at the same time
 - Asynchronous Many Tasking (AMT)



What is Kokkos?



Kokkos
performance portability for C++ applications



Cornerstone for performance portability across next generation HPC architectures at multiple DOE laboratories, and other organizations.



Abstractions

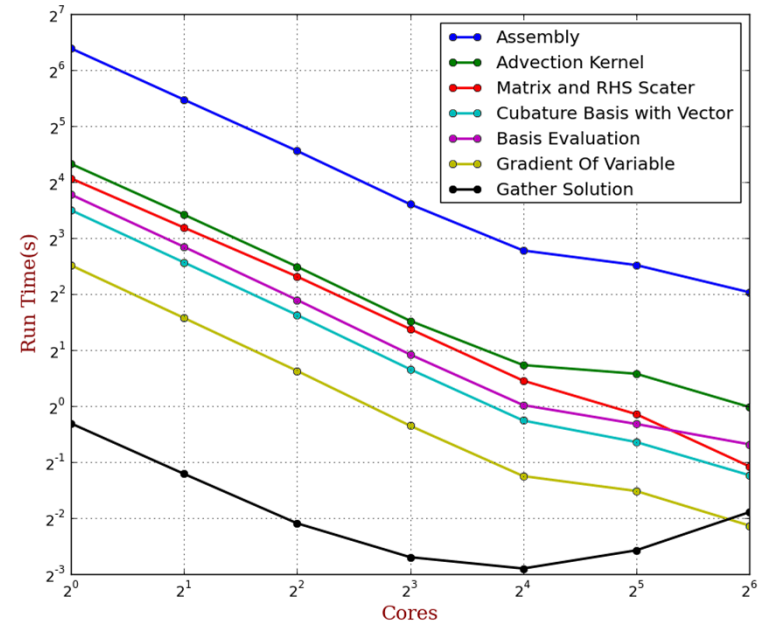
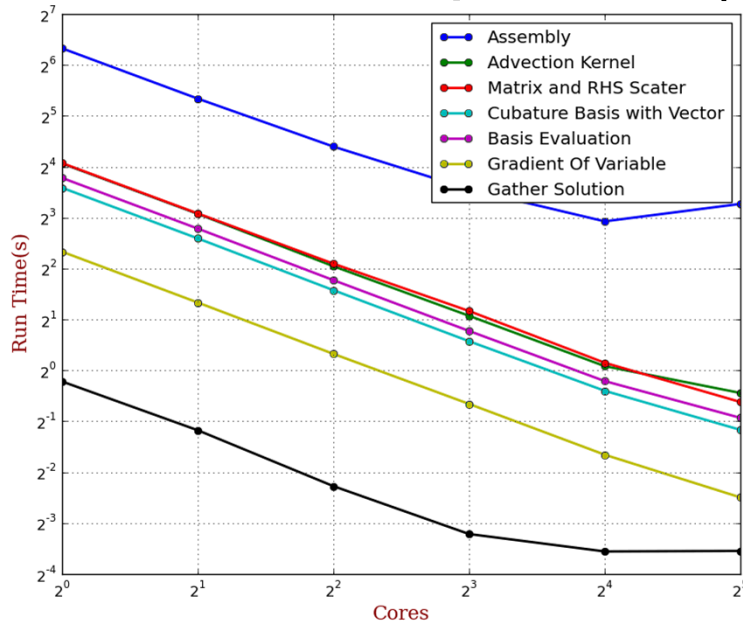
Patterns, Policies, and Spaces

- Parallel Pattern of user's computations
 - `parallel_for`, `parallel_reduce`, `parallel_scan`, task-graph, ... (*extensible*)
- Execution Policy tells *how* user computation will execute
 - Static scheduling, dynamic scheduling, thread-teams, ... (*extensible*)
- Execution Space tells *where* computations will execute
 - Which cores, numa region, GPU, ... (*extensible*)
- Memory Space tells *where* user data resides
 - Host memory, GPU memory, high bandwidth memory, ... (*extensible*)
- Layout (policy) tells *how* user array data is laid out
 - Row-major, column-major, array-of-struct, struct-of-array ... (*extensible*)
- Differentiating: Layout and Memory Space
 - Versus other programming models (OpenMP, OpenACC, ...)
 - Critical for performance portability ...

Threading

Traditional Architectures

- The gather, math routines and scatter operations were all threaded
- MPI only (left) was compared to Kokkos::OpenMP (right) on 16000 elements

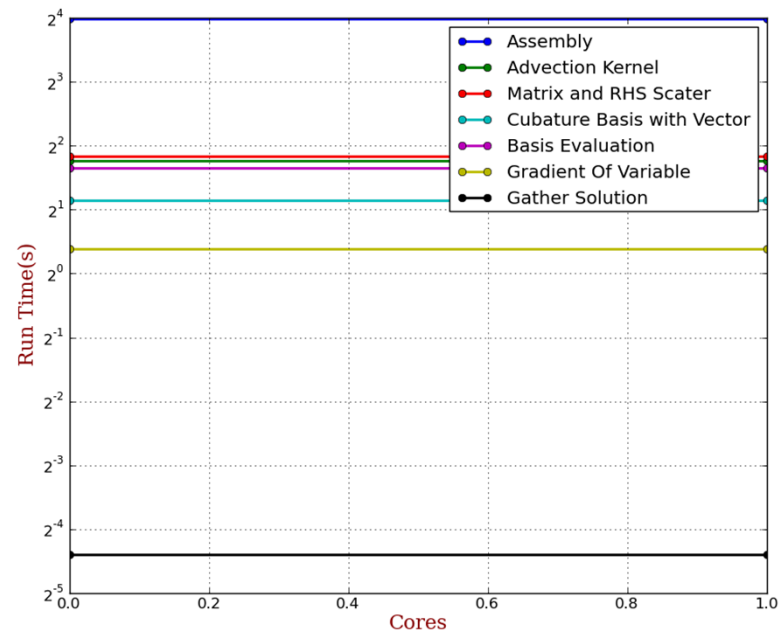
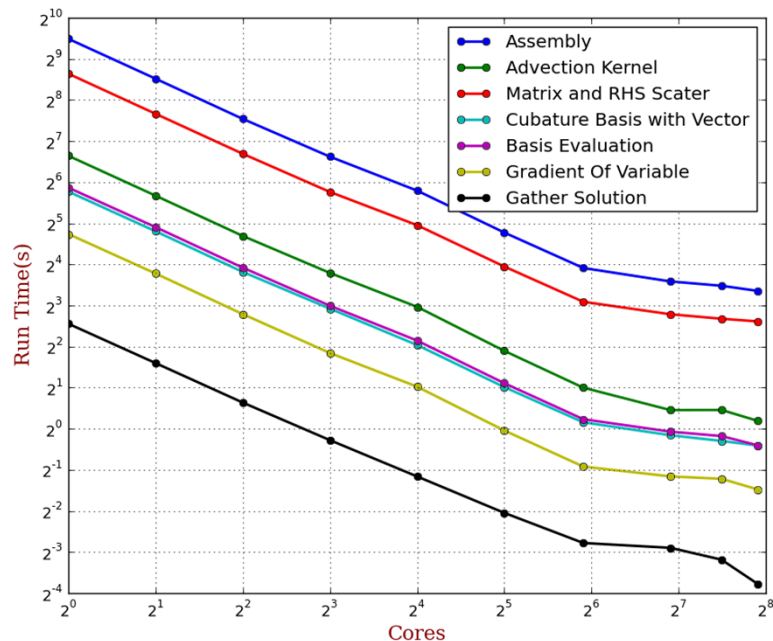


Run on a dual 16 core Haswell with hyperthreading. Left result uses one socket for 1 to 32 cores and MPI across sockets. MPI results show the average time across MPI ranks. 156k total degrees of freedom. Time is for Jacobian assembly only.

Threading

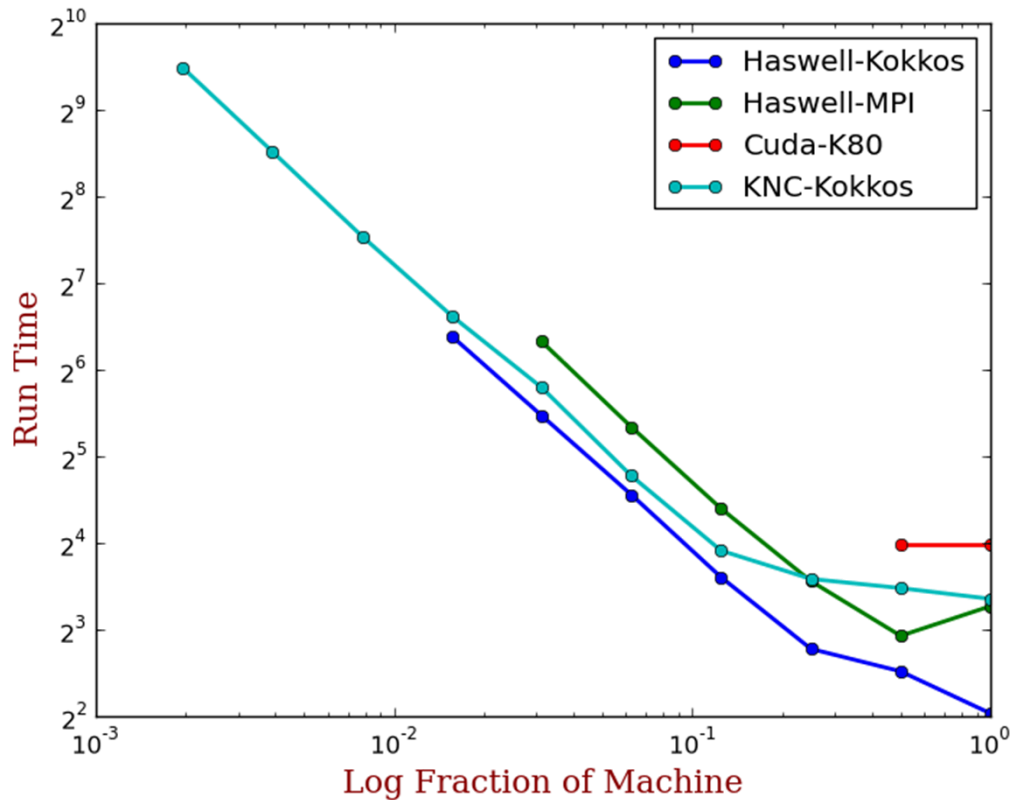
Modern Architectures

- Kokkos allows one to migrate to new machines and backends
 - Only recompilation is required to move to Intel's Knights Corner (left) or Nvidia K80 via Cuda (right)



Summary

Threading Results





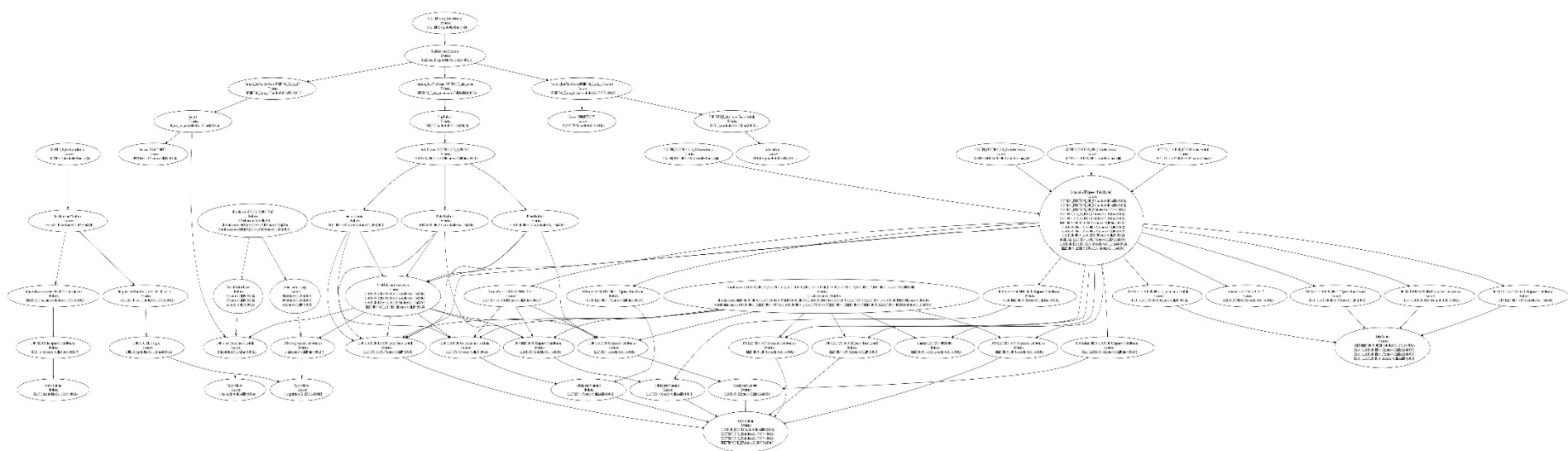
Beyond Threading

- Threading accelerates each individual math kernel
 - Eventually there is insufficient work in a single kernel to use all the threads efficiently
- Task parallelism
 - Chunks of threads solve different kernels
 - The wider the graph, the more task parallelism
 - Asynchronous Many Task (AMT) is an active research area in scheduling these tasks
 - HPX, DHARMA, Charm++, Legion, ...

Task Parallelism

Single Fluid

Real DAG much more complex than the previous toy example



Theoretical speed up for this DAG

	1 Thread/Kernel	8 Threads/Kernel	16 Threads/Kernel
Jacobian	3.5	4.5	4.9
Residual	3.4	3.4	3.5



Summary

Next Steps

- Assembly performance can be improved with both threading and task parallelism
 - Threading shows up-to 64x speed up
 - Need to demonstrate these two approaches together
- Cuda parallelism performance lower than expected
 - Not enough parallelism per workset
 - Need bigger workset, however, limited memory
 - Need to implement hierarchical parallelism
- Kokkos makes thread parallelism easier
 - Need to incorporate Kokkos further