The Scalable Data Management, Analysis, and Visualization Institute http://sdav-scidac.org

VTK-m: Accelerating the Visualization Toolkit for **Multi-core and Many-core Architectures**

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VTK-m Goals

- A single place for the visualization community to collaborate, contribute, and leverage massively threaded algorithms
- Reduce the challenges of writing highly concurrent algorithms by

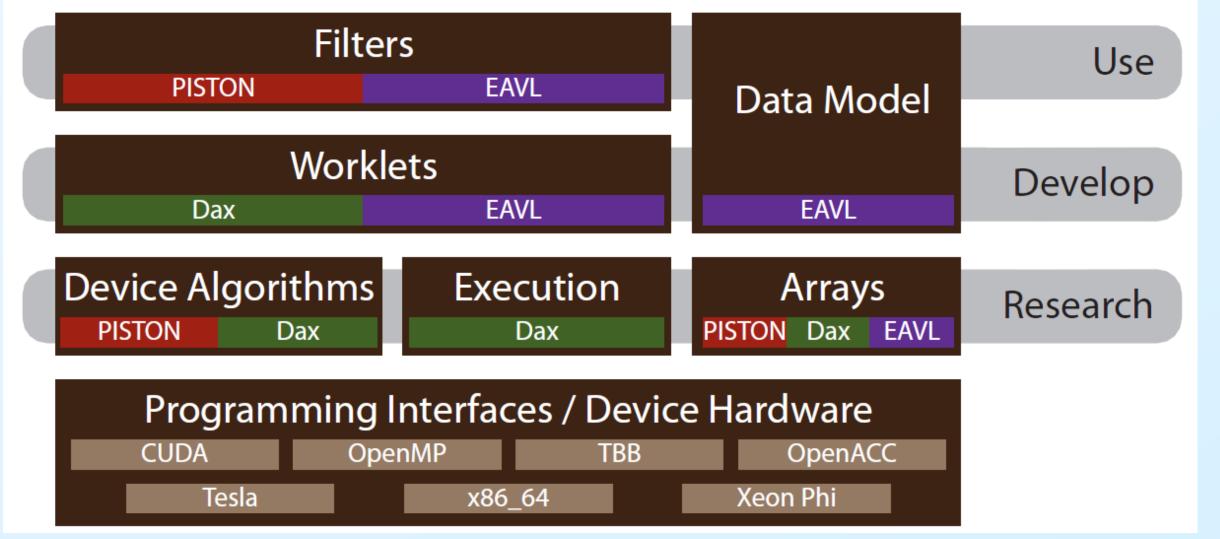
Cosmology Applications

- Halo finding and halo center finding algorithms were written using PISTON, one of VTK-m's constituent projects
- On Titan, this enabled centers to be found on the GPU ~50x

Hardware-Agnostic Ray Tracing

- VTK-m's hardware-agnostic approach gives comparable performance to hardware-specific approaches
- Since VTK-m is implemented in a hardware-agnostic way, we

- using data parallel algorithms
- Make it easier for simulation codes to take advantage these parallel visualization and analysis tasks on a wide range of current and next-generation hardware
- Unify efforts in this area from Sandia (Dax), Oak Ridge (EAVL), and Los Alamos (PISTON)

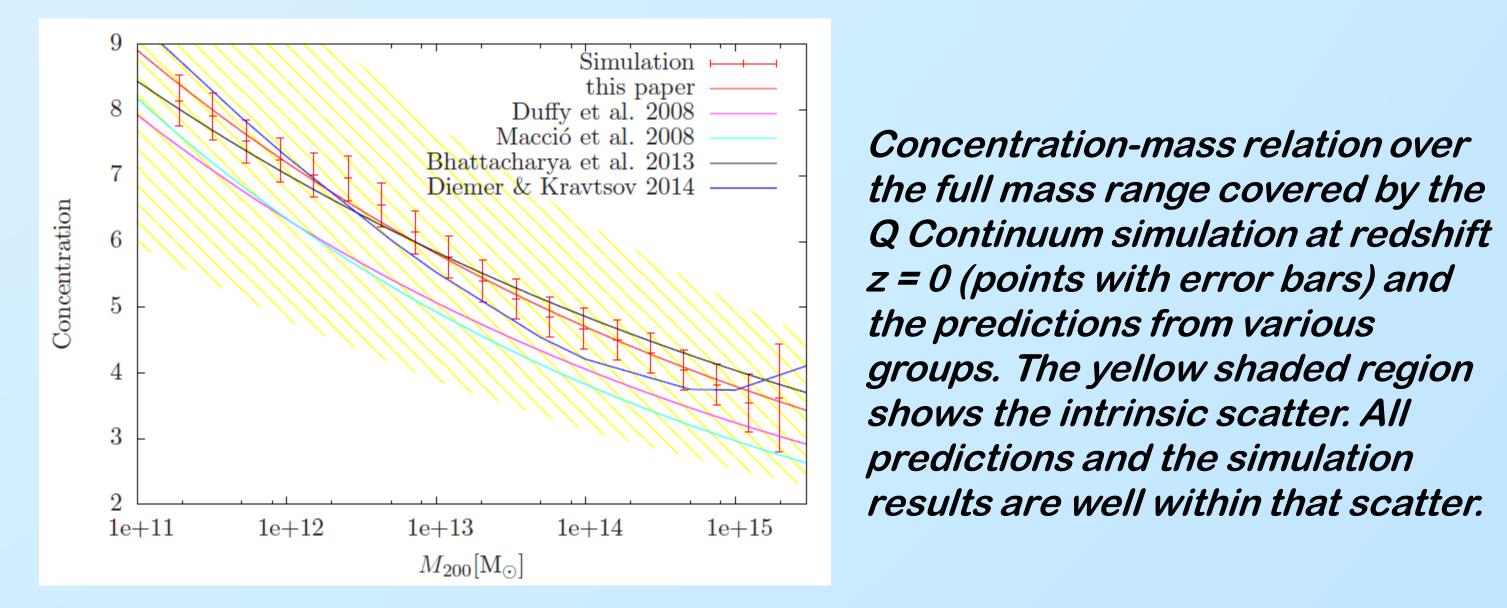


VTK-m infrastructure and use cases, with contributions from Dax, EAVL, and PISTON predecessor projects

VTK-m Status

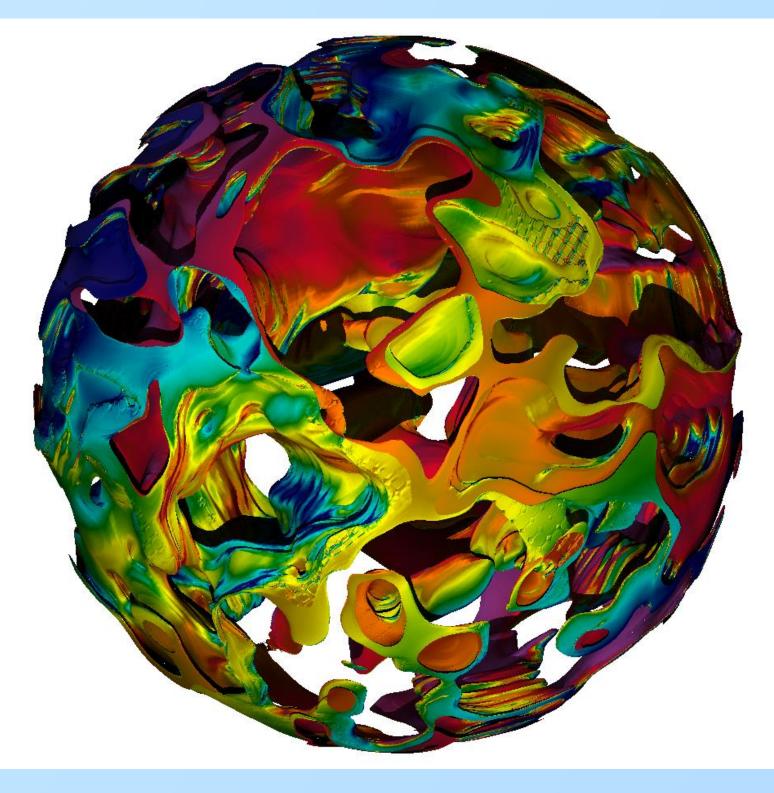
Project infrastructure

- faster than using the pre-existing algorithms on the CPU (with one rank per node)
- This work allowed halo analysis to be completed on all time steps of a very large 8192³ particle data set across 16,384 nodes on Titan for which analysis using the existing CPU algorithms was not feasible
- The portability of VTK-m allowed us to run the same code on an Intel Xeon Phi
- This is the first time that the c-M relation has been measured from a single simulation volume over such an extended mass range
- To appear in the Astrophysical Journal: "The Q Continuum Simulation: Harnessing the Power of GPU Accelerated Supercomputers".



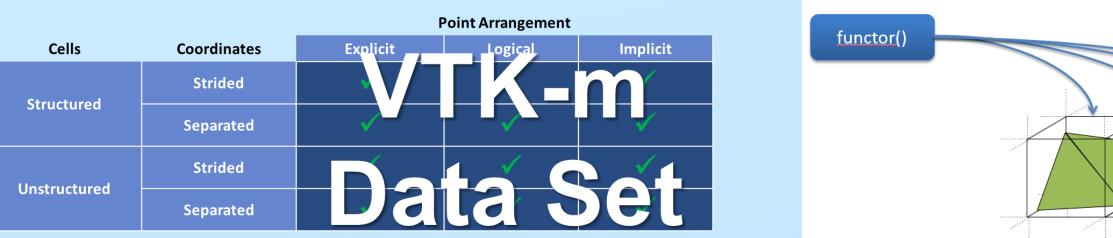
wanted to understand the corresponding sacrifice in performance

- We implemented a ray-traced renderer, which is computationally intensive and uses many unstructured memory accesses
- We then compared VTK-m's performance to NVIDIA's OptiX and Intel's Embree, two "guaranteed not to exceed" ray-tracing standards that are developed by teams of professionals
- Our study found that VTK-m performance was always within a factor of two of industry standards, and even outperformed them in some cases
- We concluded that VTK-m hardware-agnostic approach is viable our single implementation performed comparably to multiple hardware-specific implementations



Ray-traced rendering of 6.2M triangles generated from SPECFEM3D. The data represents wave speed perturbations measured by seismograms and was provided by Oak

- Code repository: <u>https://gitlab.kitware.com/vtk/vtk-m</u>
- Project webpage: http://m.vtk.org
- Features
 - Core Types
 - Statically Typed Arrays
 - Dynamically Typed Arrays
 - Device Interface (Serial, CUDA, TBB; OpenMP in progress)
 - Field and Topology Worklet and Dispatcher
- Data Model
 - Allows clients to construct data sets from cell and point arrangements that exactly match their original data
 - In effect, this allows for hybrid and novel mesh types
- Filters
- Isosurface for structured grids
- Statistical filters (histograms, moments, etc.)
- In development: stream lines, stream surfaces, tetrahedralization



In-situ Applications

- Tightly coupled in-situ with EAVL, one of VTK-m's constituent projects
 - Efficient in-situ visualization and analysis
 - Light weight, zero-dependency library
 - Zero-copy references to host simulation
 - Heterogeneous memory support for accelerators
 - Flexible data model supports non-physical data types
 - Example: scientific and performance visualization, tightly coupled EAVL with SciDAC Xolotl plasma surface simulation
- Loosely coupled in-situ with EAVL
 - Application de-coupled from visualization using ADIOS and Data Spaces
 - EAVL plug-in reads data from staging nodes System nodes running EAVL perform visualization operations and rendering Example: field and particle data, EAVL in-situ with XGC SciDAC simulation via ADIOS and Data Spaces

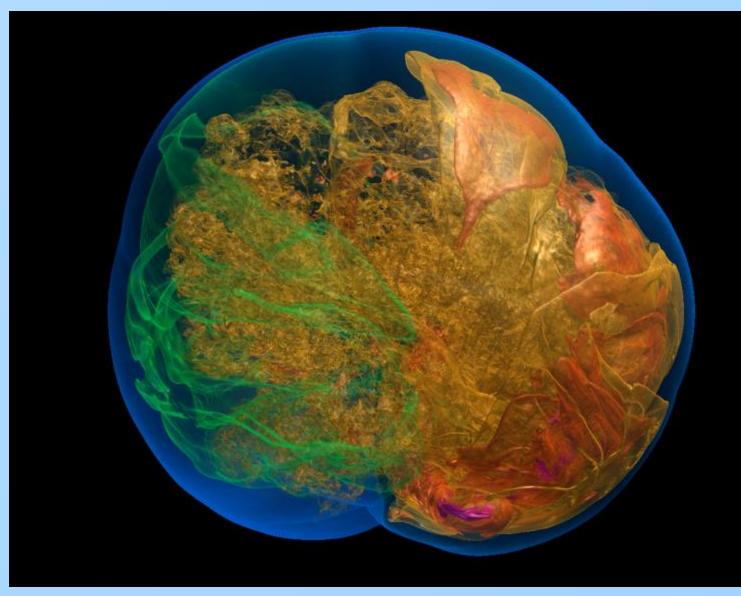
predictions and the simulation results are well within that scatter.

Ridge National Laboratory.

Advanced Visualization Usability Study

- Implementation of both ray-casting and cell projection volume rendering algorithms using Dax, one of VTK-m's constituent projects
- Complied for CUDA, OpenMP, and Intel's Thread Building Blocks
- Comparative performance study on NVIDIA Titan X GPU, Intel Xeon, and Intel Xeon Phi
- VTK-m implementation in progress

Volume rendering of type la supernova simulation data set using ray-casting. Cell projection implementation using data parallel primitives renders comparable image in near sub-

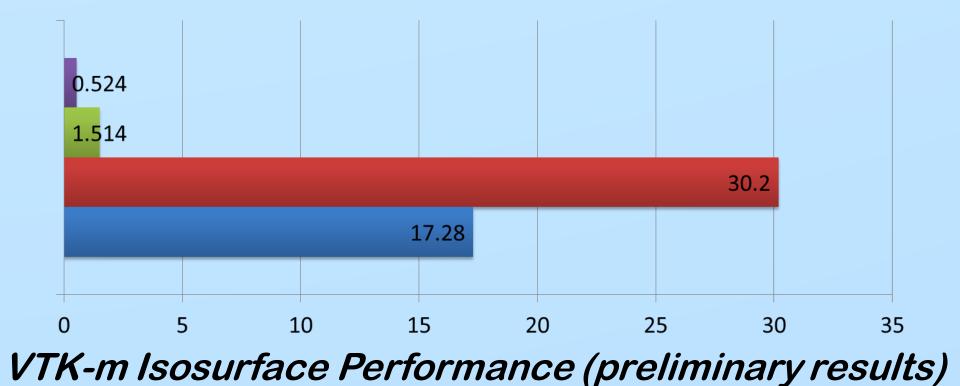




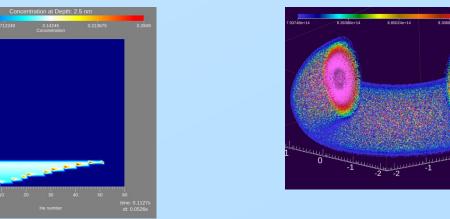
VTK-m Data Model

Marching Cubes

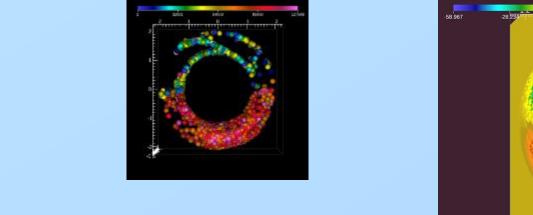
■ VTK-m Cuda [No Transfer] ■ VTK-m Cuda ■ VTK-m Serial ■ VTK Serial



Functional programming paradigm



EAVL in-situ with Xolotl



EAVL in-situ with XGC

second times.

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