

ALPINE: In Situ Visualization, Analysis, and Infrastructure for ECP Science

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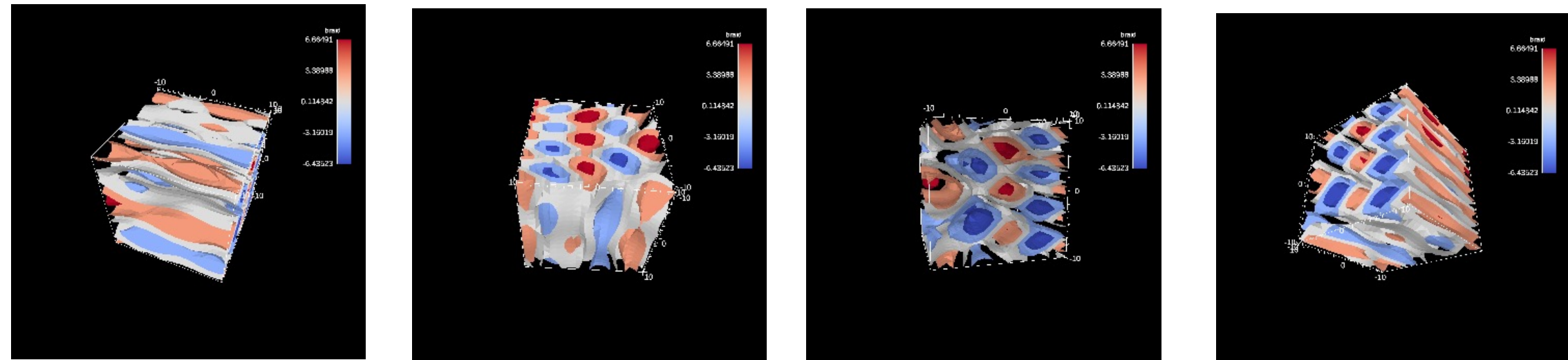
Overview

ALPINE is leveraging current analysis and visualization software while developing new capabilities to deliver **in situ visualization, infrastructure, and data analysis to ECP Applications**. The project focuses on these main development areas:

- Deliver **visualization and analysis algorithms** that are critical for ECP Applications as the dominant analysis paradigm shifts from *post hoc* to *in situ*.
- Deliver an exascale-capable **infrastructure** for the development of in situ algorithms and deployment into existing applications, libraries, and tools.
- **Integrate** our algorithms and infrastructure with ECP Application software and workflows.
- **Integrate** ECP Software Technologies into our infrastructure to meet the needs of ECP Applications.

Ongoing Infrastructure, Algorithm Development & Integration Efforts

Optimal Viewpoint Metrics based on data properties can automate in situ visualization decisions to only capture interesting views. Metrics left to right: data entropy, depth entropy, max depth, projected area.



NekRS ↔ Ascent

Ascent-NekRS initial testing on low-order, linear elements. Using the Ascent-NekRS integration, volume-rendered flow-field images were generated at multiple time-steps for a simple test case, *couette* flow, which has an analytical solution.

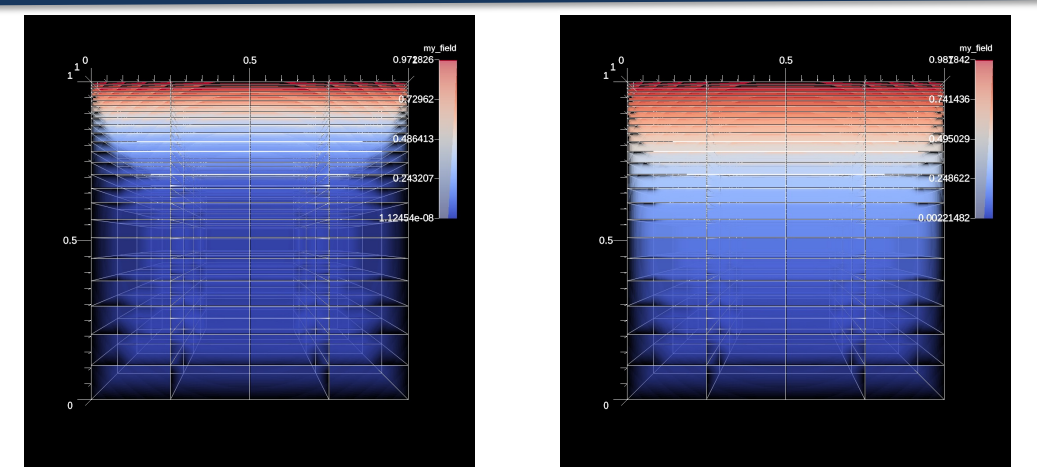
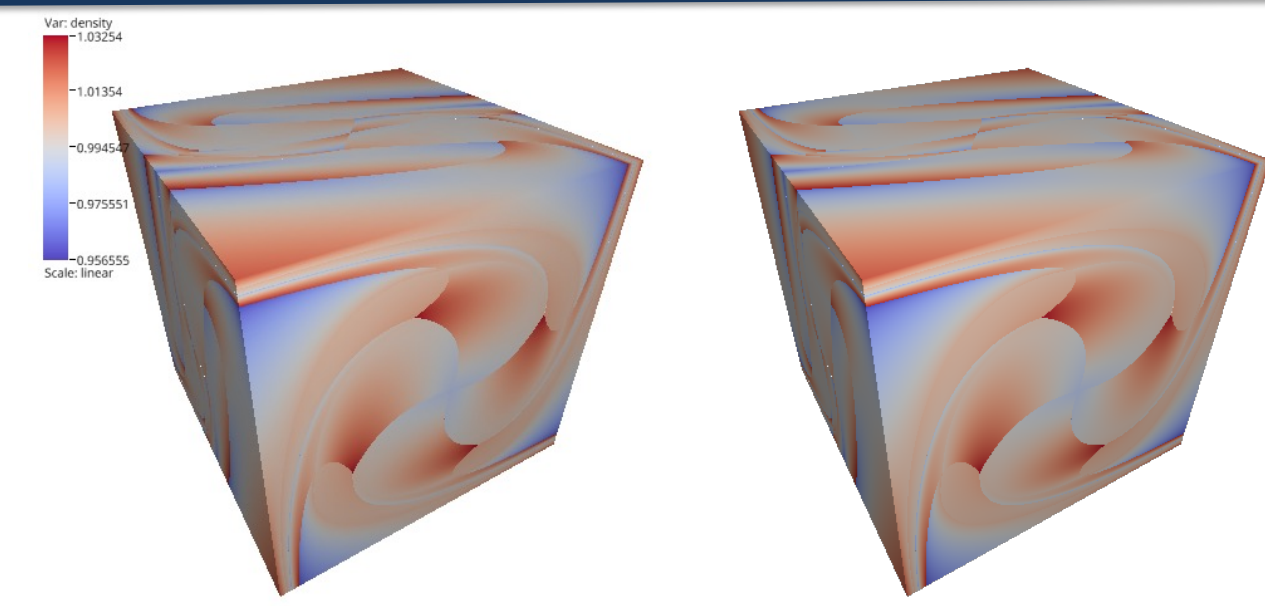


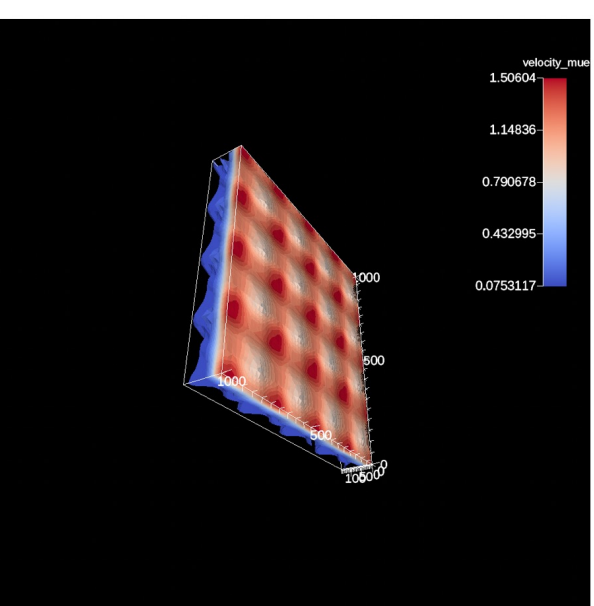
Image rendering of a simple, couette flow at early (t=0.1) and later time step (t=1.0). The field shown represents velocity magnitude.

Task-based hierarchical feature extraction algorithm based on segmented merge trees. The algorithm is implemented using a multi-runtime abstraction layer, BabelFlow, which can be used to execute arbitrary analysis and visualization dataflows using different task-based runtime systems.

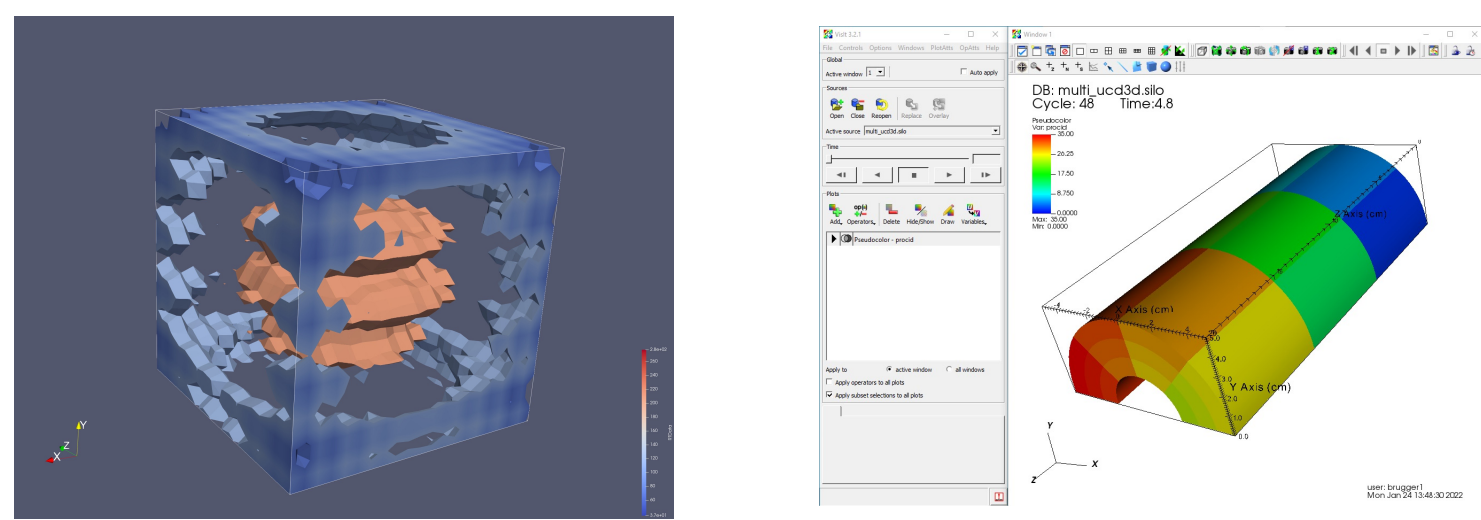


Left: Rendered/composited via DevilRay & VTK-h. Right: Rendered/composited via DevilRay & BabelFlow's RadlX algorithm.

Ascent ↔ AMR-Wind Integration ported to Crusher



ALPINE Infrastructure on Frontier EAS

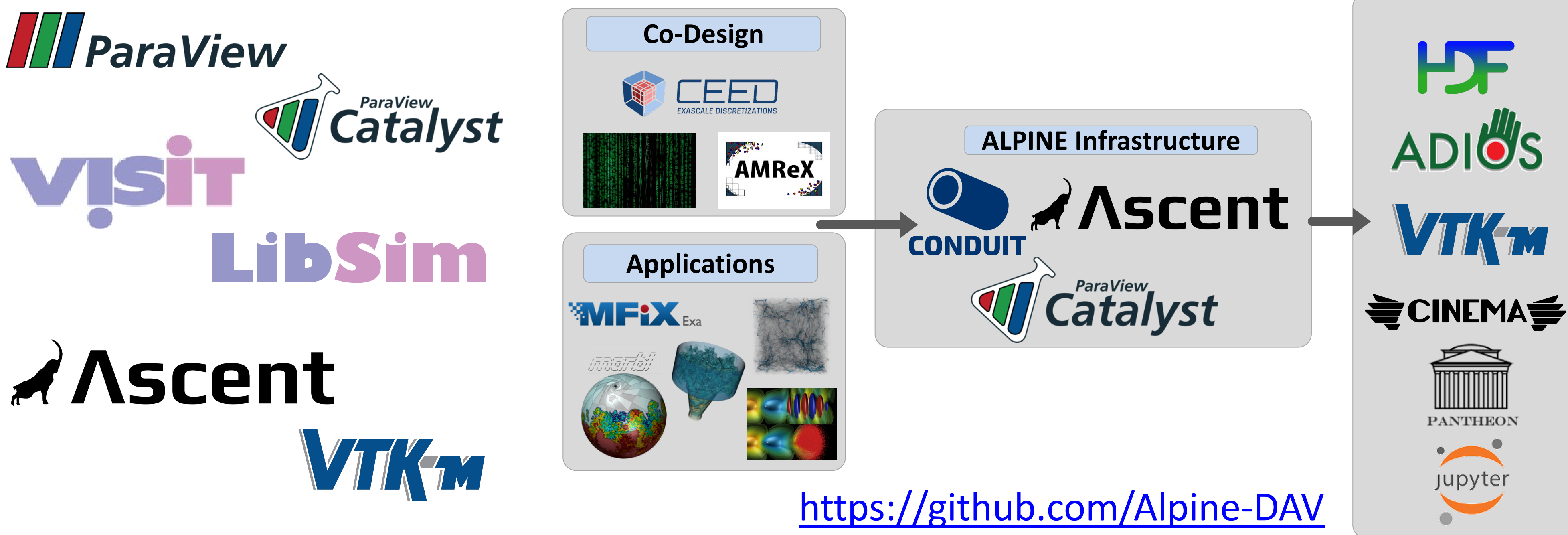


Images from ParaView & VisIt Spock porting activities. ParaView and VisIt compiled on the CPU using OSMesa. Left, a pybatch image from a rendering test. Right: a VisIt client-server demo of a parallel run. Ascent has been ported to Crusher GPUs.

ALPINE LCF Teams: Delivering on Exascale platforms

ALPINE teams at Oak Ridge and Argonne are providing support for ALPINE infrastructure and algorithms on early access systems to meet the data and visualization needs of ECP applications.

Software Stack and Infrastructure



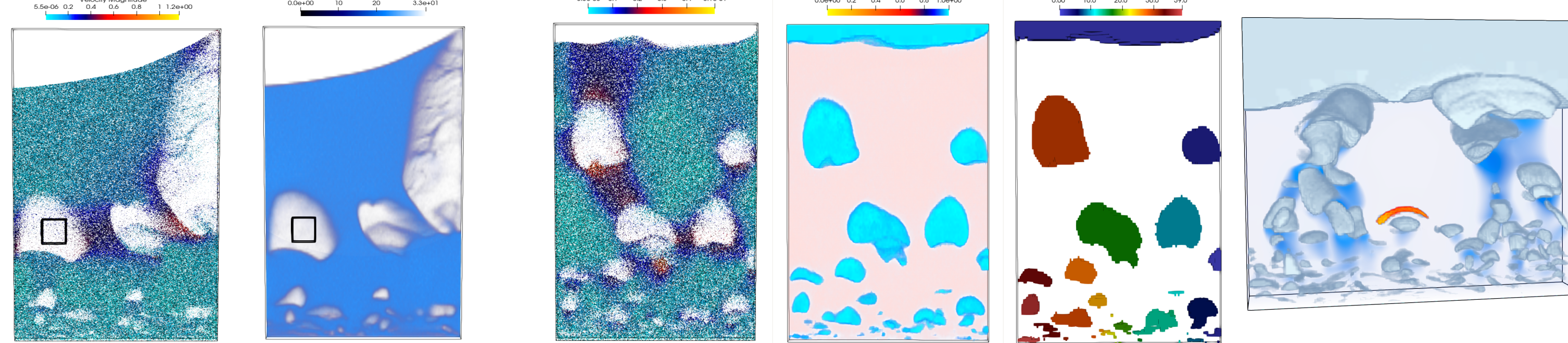
<https://github.com/Alpine-DAV>

Example Integrations & Workflows



Goal: in situ data reduction & feature detection with post hoc interactive analysis via Cinema

In situ **statistical feature detection** detects features in particle data sets using statistical data modeling and probabilistic similarity measures.



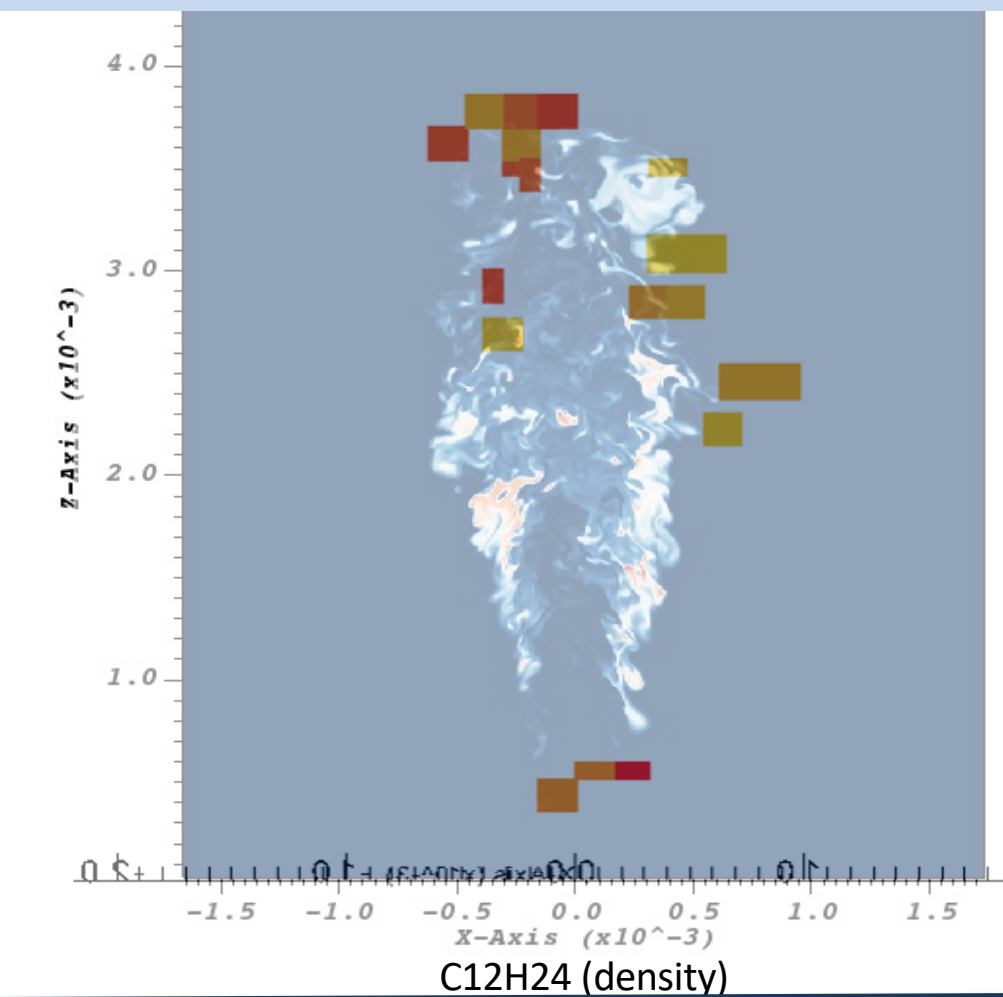
Above: Figures illustrate process to identify "voids" or bubbles in MFiX-Exa particle data. L to R: (1) Void selection: box shows region of interest with low particle density. (2) Voids as seen in density field. (3) Voids in particle field. (4) Feature similarity field - how similar to region of interest. (5) Voids are isolated by thresholding for analysis and characterization. (6) Production visualization.

Combustion-Pele



Goal: in situ anomaly detection to trigger downstream analysis for auto-ignition events

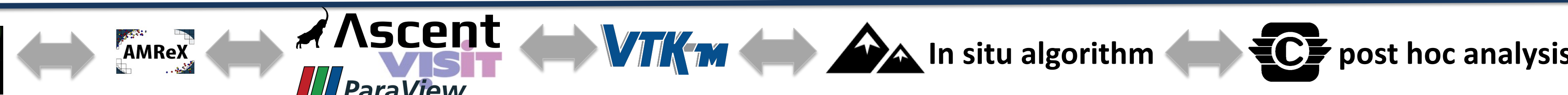
ALPINE, ExaLearn and Pele collaboration: AI/ML techniques independently developed in ExaLearn are being deployed in situ with Ascent technology from ALPINE to detect auto-ignition regions in a PeleLM simulation. Demonstrated: GPU-scalable in situ detection of auto-ignition regions using co-kurtosis tensor-based anomaly detection.



Left: Simulation snapshot of an intermediate species overlaid with AMR boxes flagged by the anomaly detection algorithm. Right: production visualization via Ascent. Image left courtesy of M. Rieth, J. Chen (Pele), M. Arienti, M. Larsen, J. Bennett (ALPINE), H. Kolla (ExaLearn). Image right courtesy of M. Larsen (ALPINE).

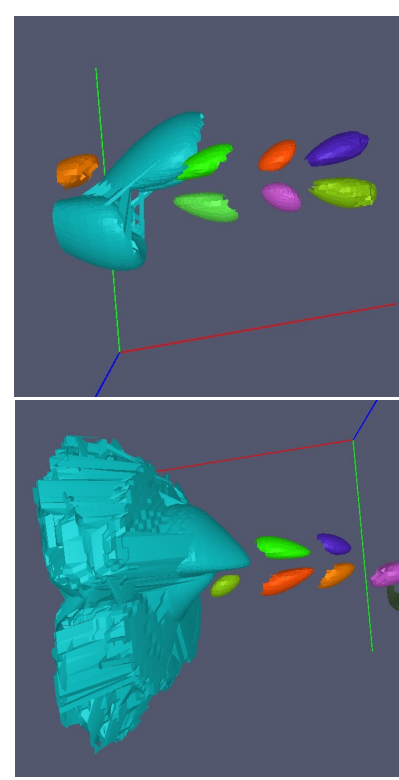


WarpX

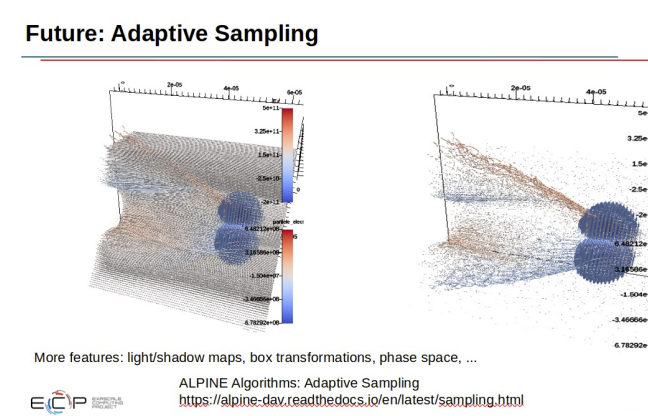
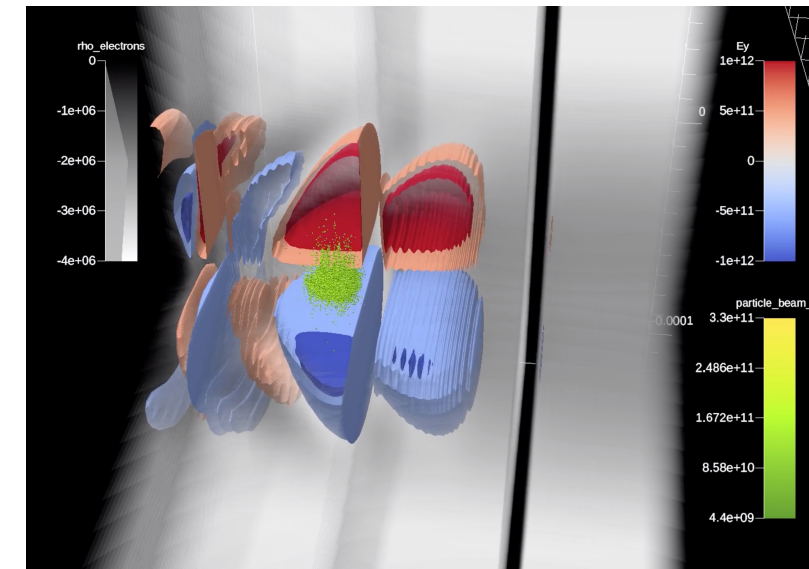


Goal: Exploratory visualization and analysis to understand physics of plasma-based particle accelerators

Topological analysis is used to identify most relevant contours & create isosurface visualizations in situ; saving resulting images for post hoc analysis. Images are saved to a Cinema DB in a format that supports arbitrary combination of contours during post hoc visualization. Right: Most relevant contours in WarpX simulation selected using two importance measures: persistence, volume.



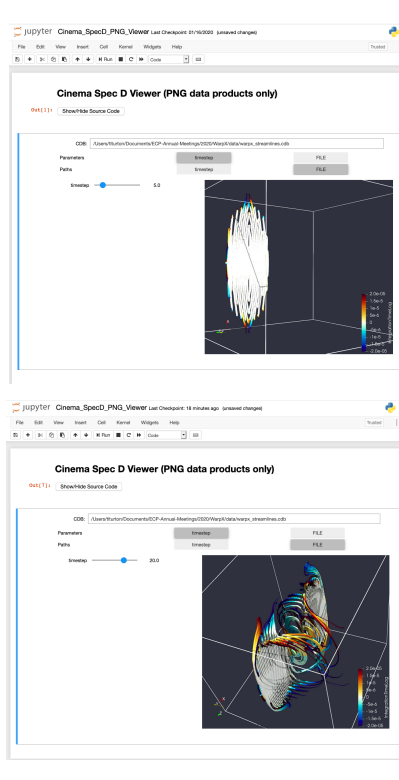
Below: visualization of particles and fields produced in situ can be used to generate movies showing complex physics behavior over time.



Left top: Sampling on transverse momentum reveals particles behaving unexpectedly. Image courtesy of A. Huebl (WarpX)

Left bottom: VTK-m WarpX-specific particle advection algorithm visualized post hoc in VisIt. Image courtesy of H. Childs, A. Yenpure (VTK-m)

Right: Post hoc analysis workflow via ParaView to visualize EM field streamlines in Cinema Jupyter notebook. Image courtesy of R. Bujack, L. Turton, DH Rogers (Cinema)

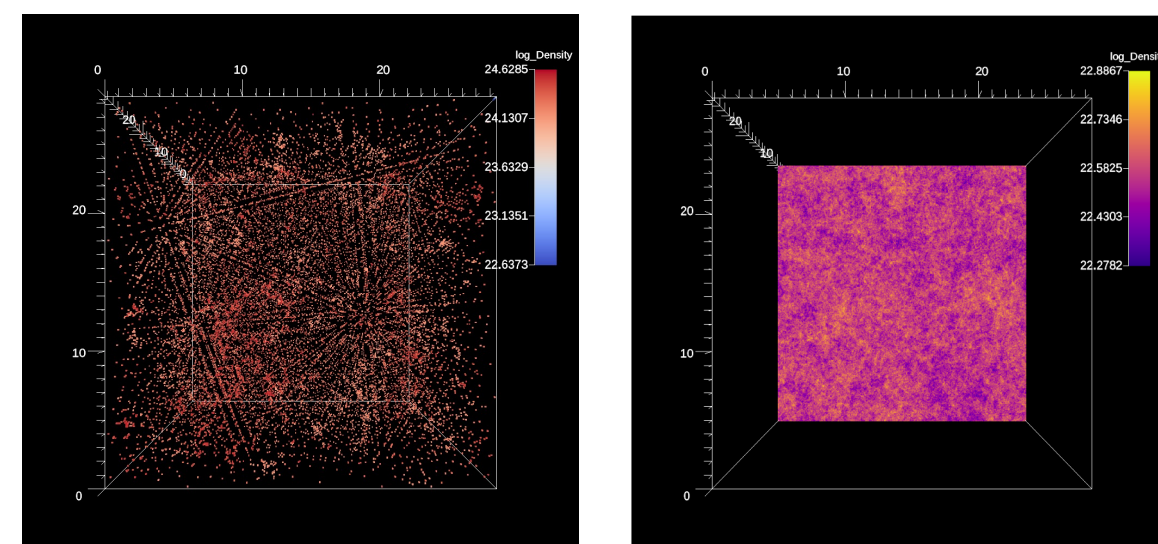


Nyx

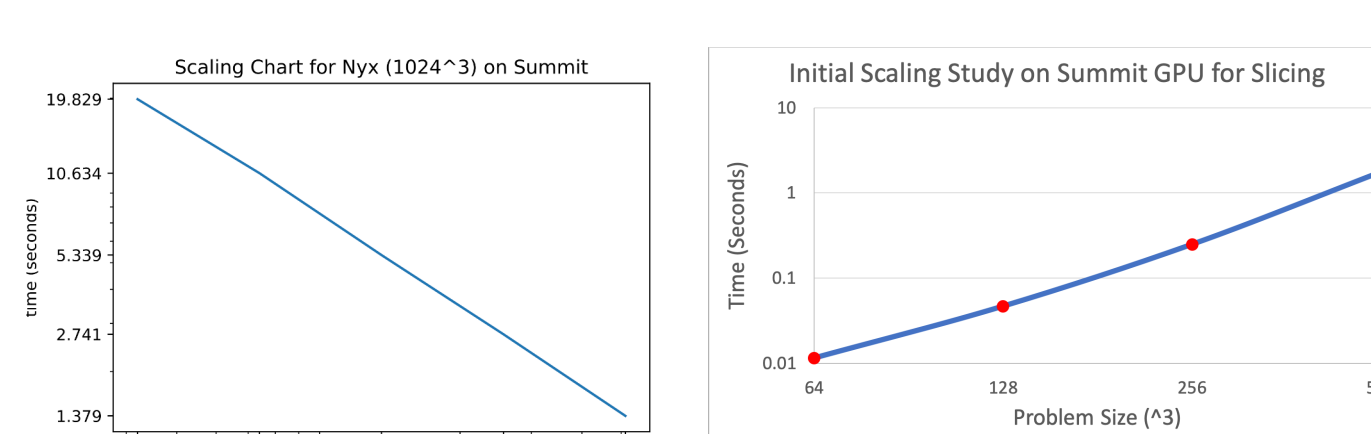


Goal: In situ data reduction and slice visualization with post hoc reconstruction

Data-driven sampling enables probabilistic identification of interesting regions in the data automatically, prioritizing important regions. Applied in situ to Nyx, important halo regions are preserved.



Left: data-driven sampling applied to Nyx data in situ. Right: slicing algorithm produces in situ visualization for post hoc exploration and validation.



The Nyx ↔ Ascent ↔ sampling/slicing pipelines have been run on Summit GPUs with excellent scaling results. Left: sampling algorithm, right: slicing algorithm.