

DIY Applications

Tom Peterka

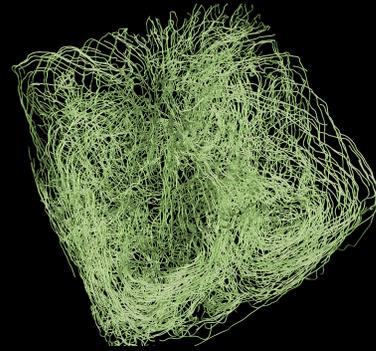
tpeterka@mcs.anl.gov

Mathematics and Computer Science Division

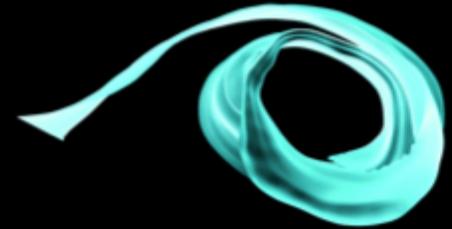
Abstractions Matter: Blocks, not Tasks

- Block = unit of decomposition
- Block size, shape can be configured
 - From coarse to fine
 - Regular, adaptive, KD-tree
- Block placement is flexible, dynamic
 - Blocks per task
 - Tasks per block
 - Memory / storage hierarchy
- Data is first-class citizen
 - Separate operations per block
 - Thread safety

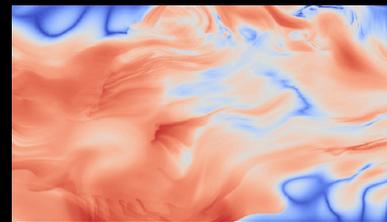
Parallel data analysis consists of decomposing a problem into blocks, operating on them, and communicating between them.



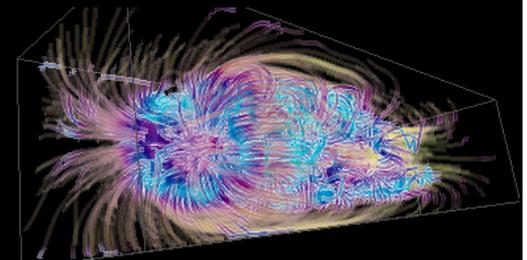
Streamlines and pathlines



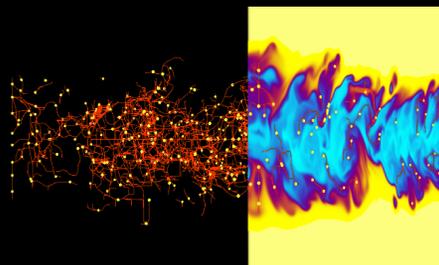
Stream surfaces



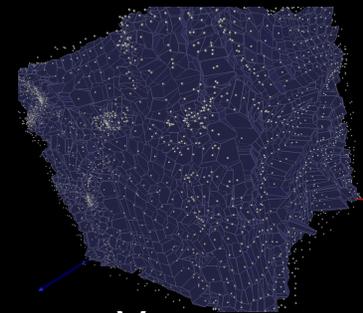
FTLE



Information entropy



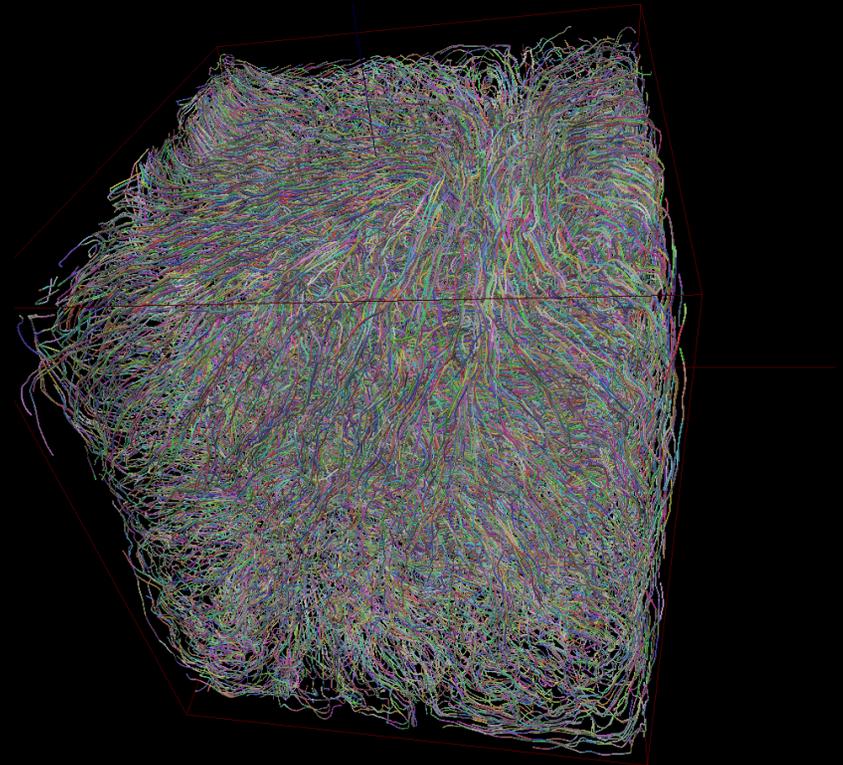
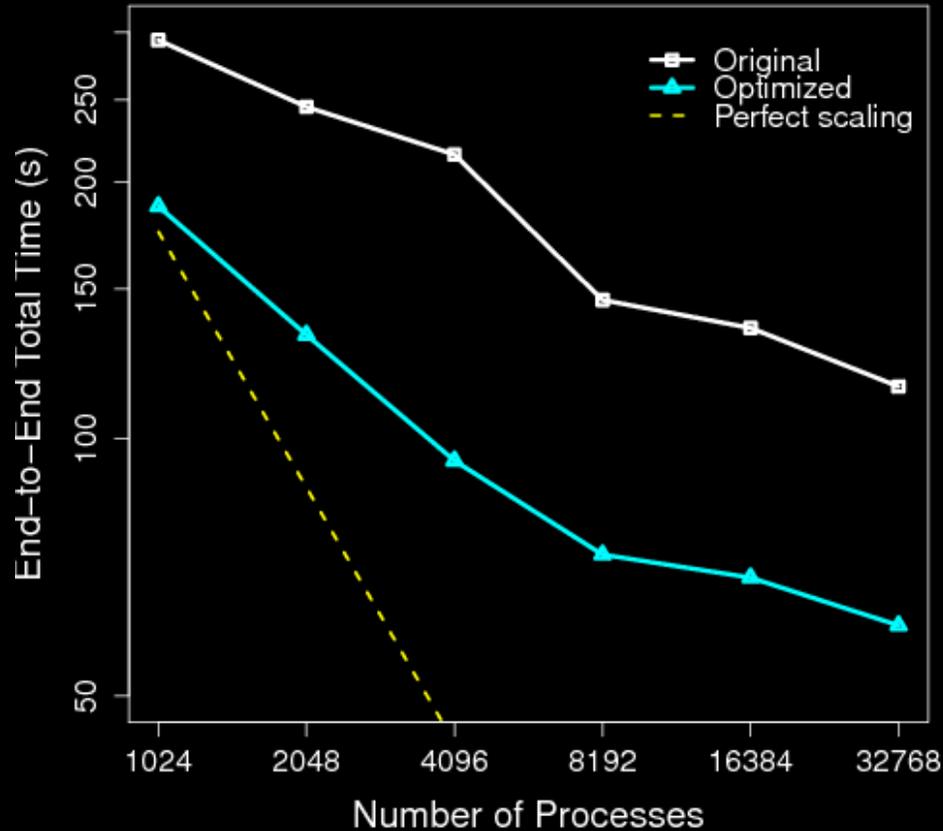
Morse-Smale complex



Voronoi Tessellation

Particle Tracing Streamlines and Pathlines

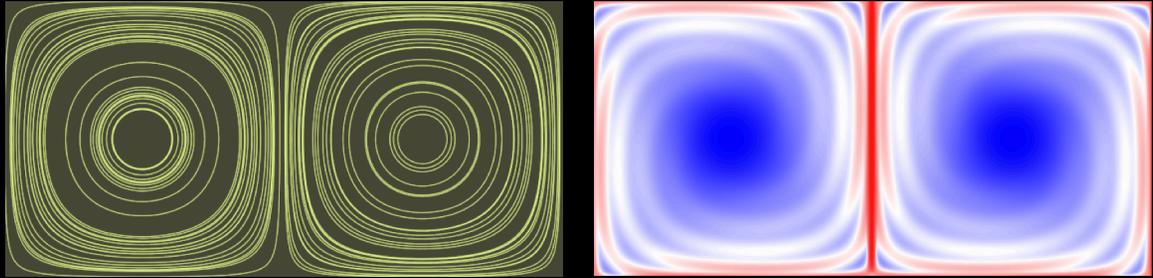
Strong Scaling



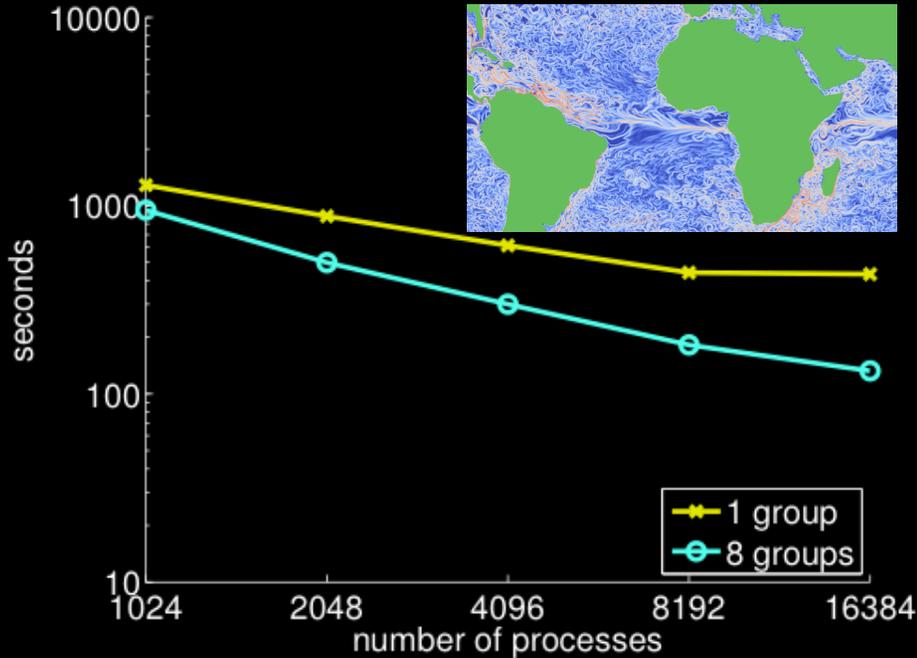
Particle tracing of $\frac{1}{4}$ million particles in a 2048^3 thermal hydraulics dataset results in strong scaling to 32K processes and an overall improvement of 2X over earlier algorithms

Lagrangian Coherent Structures from FTLE

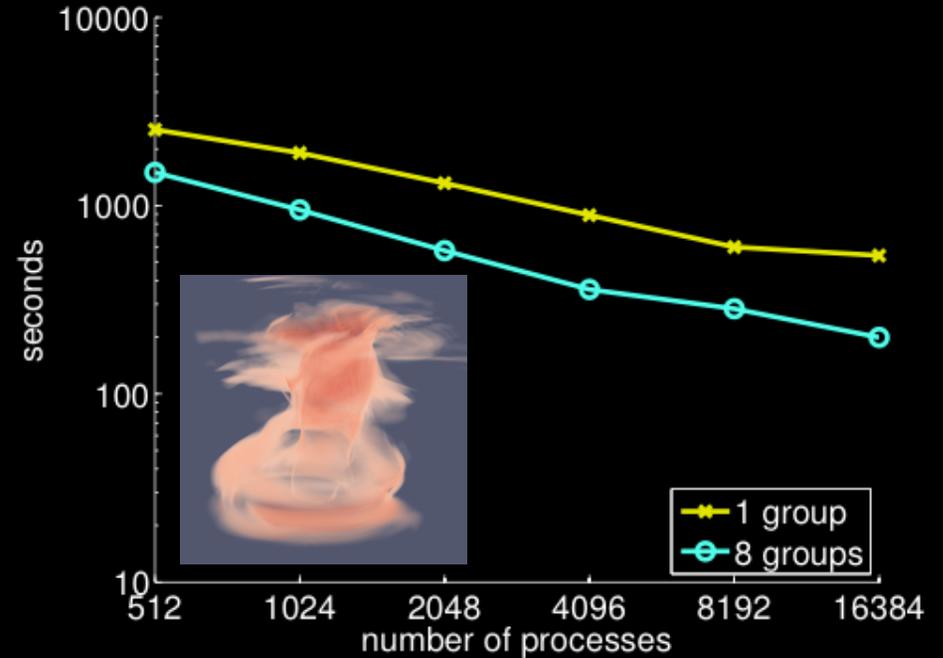
Courtesy Boonthanome Nouanesengsy



Ocean: Strong Scaling Total Time



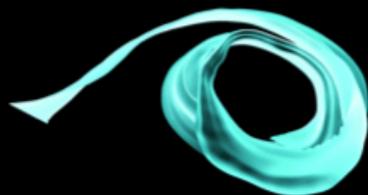
Isabel: Strong Scaling Total Time



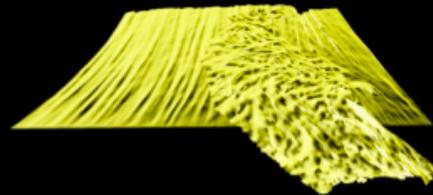
Left: Particle tracing of 288 million particles over 36 time steps in a 3600x2400x40 eddy resolving dataset. Right: 131 million particles over 48 time steps in a 500x500x100 simulation of Hurricane Isabel. Time includes I/O.

Stream Surfaces

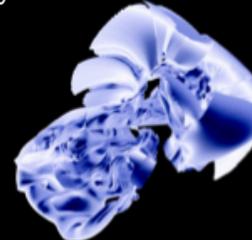
Courtesy Kewei Lu



(a) Isabel



(b) MJO

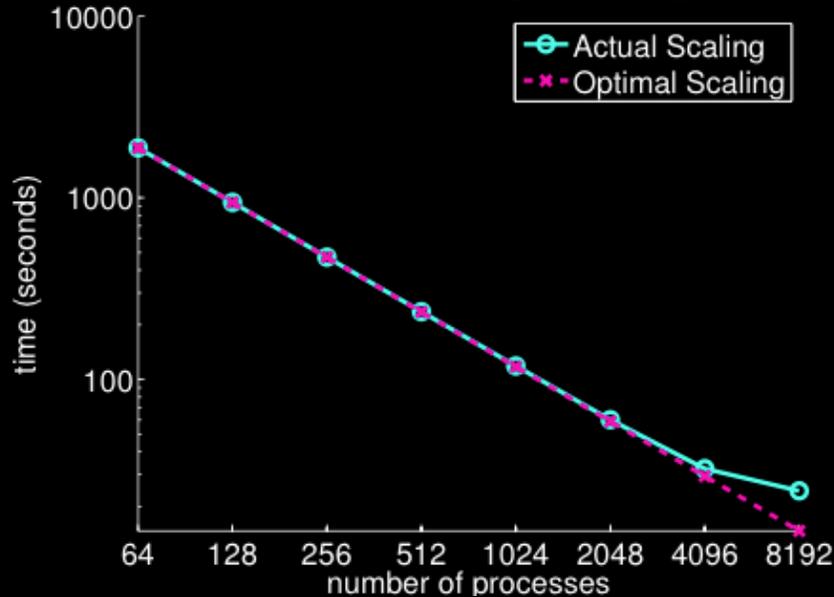


(c) Plume

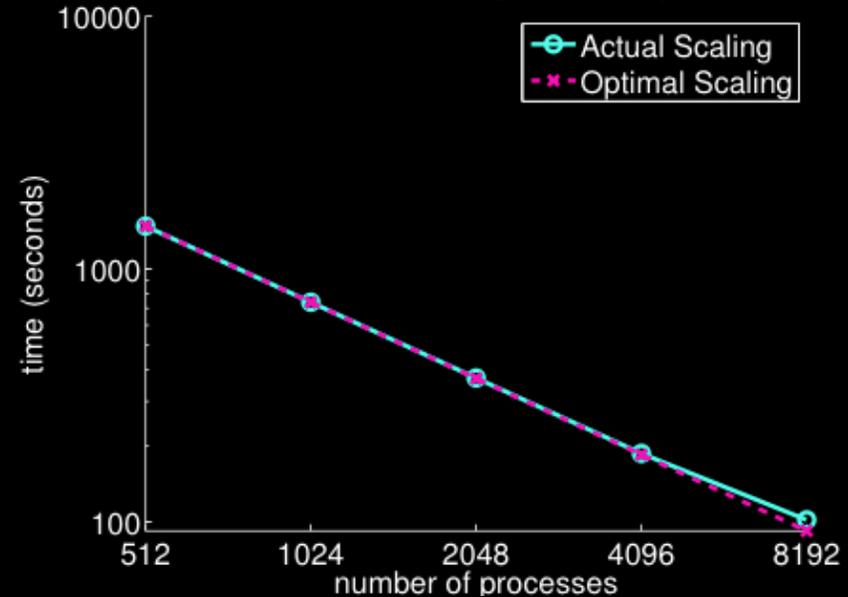


(d) Nek

Plume: Strong Scaling



Nek: Strong Scaling



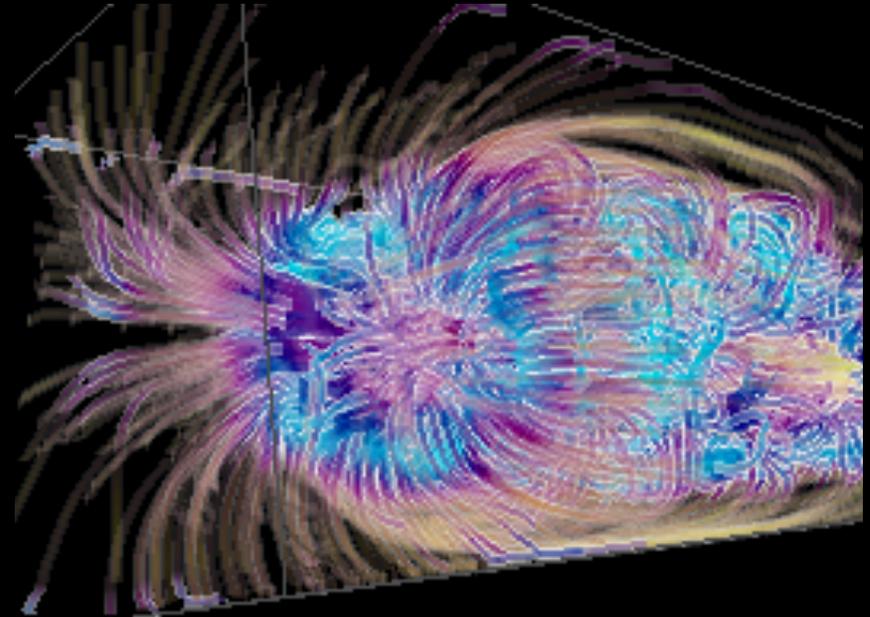
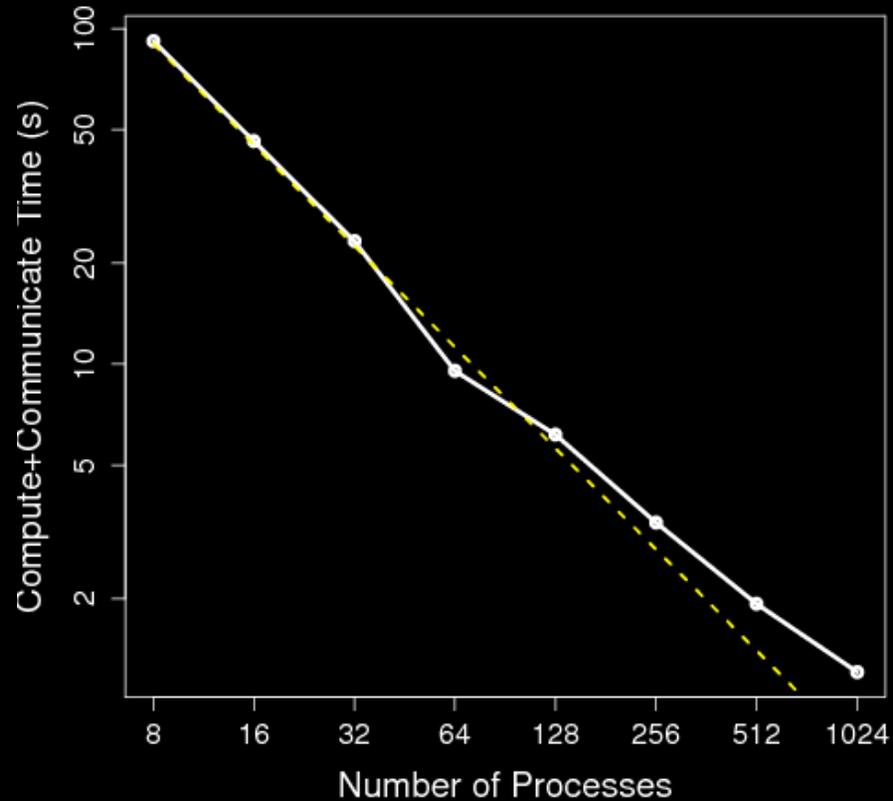
Left: 64 surfaces each seeded with 512 particles are advected in a $504 \times 504 \times 2048$ simulation of a solar flare. Right: 64 surfaces each with 2K seeds in a $2K \times 2K \times 2K$ Nek5000 thermal hydraulics simulation. Time excludes I/O.

Lu et al., Scalable Computation of Stream Surfaces on Large Scale Vector Fields, submitted to SC14.

Information Entropy

Courtesy Abon Chaudhuri

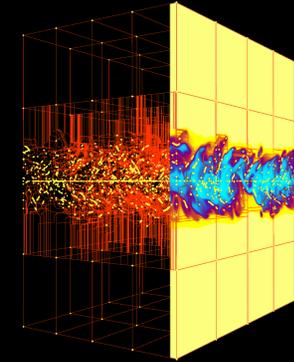
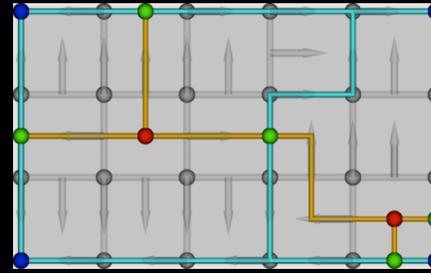
Strong Scaling



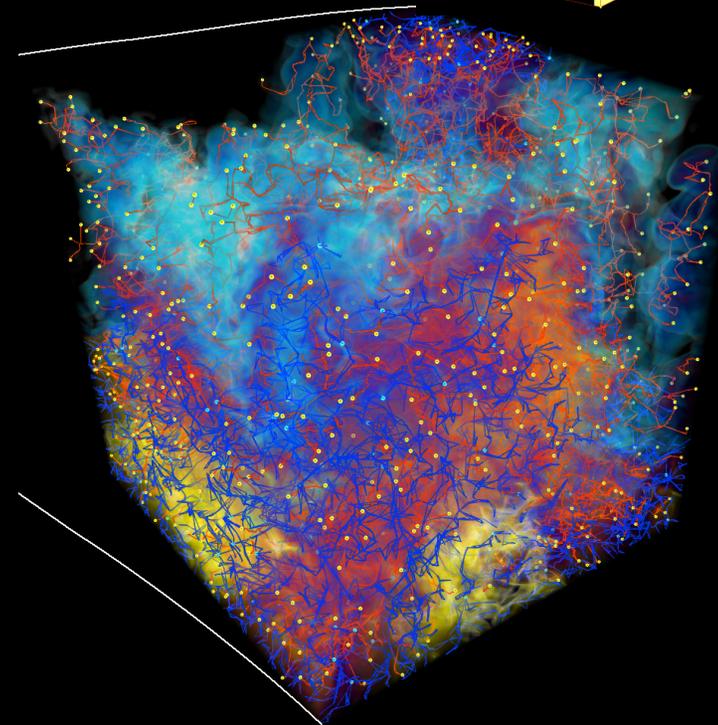
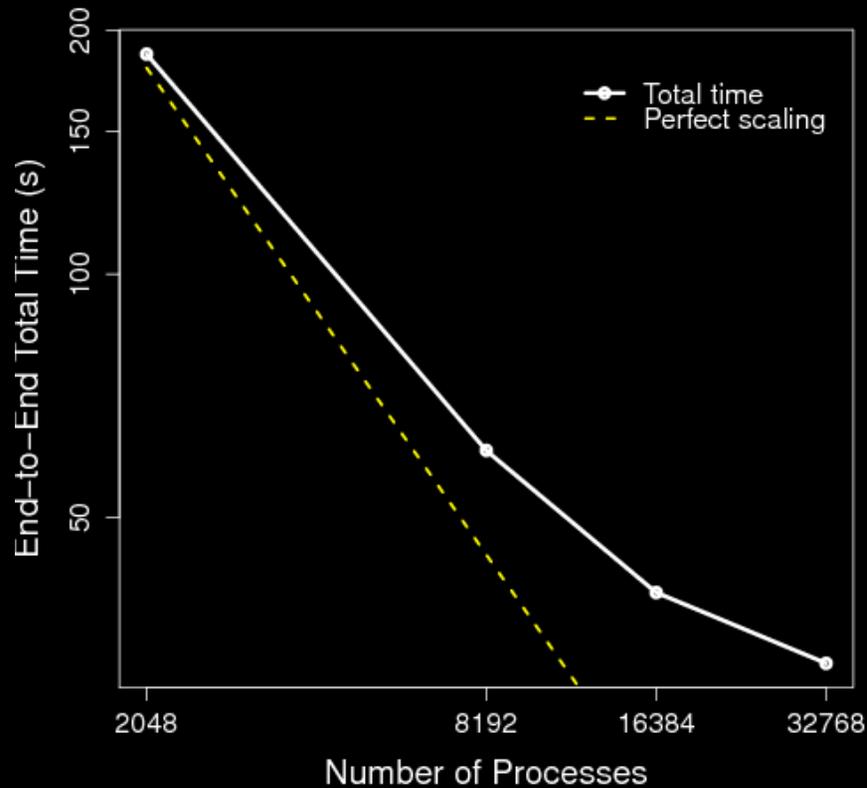
Computation of information entropy in $126 \times 126 \times 512$ solar plume dataset shows 59% strong scaling efficiency. Time excludes I/O.

Topological Analysis

Courtesy Attila Gyulassy



Strong Scaling

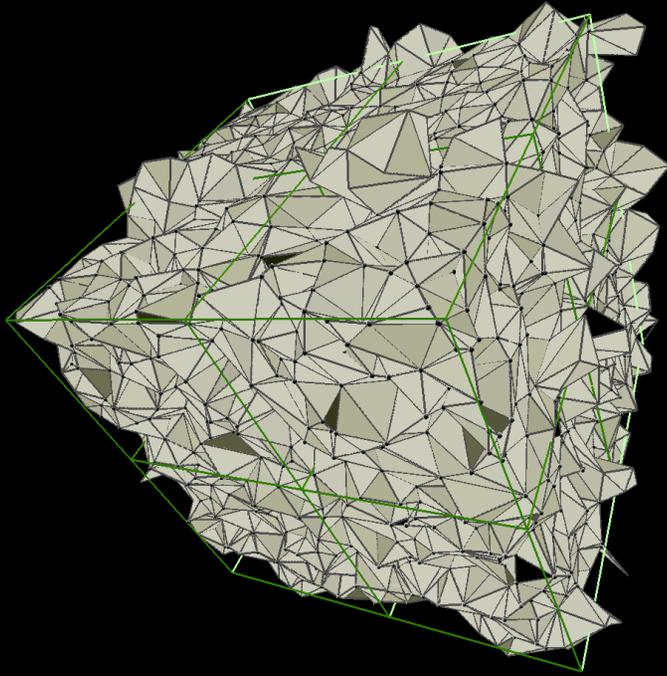


Computation of Morse-Smale complex in 1152^3 Rayleigh-Taylor instability data set results in 35% end-to-end strong scaling efficiency, including I/O.

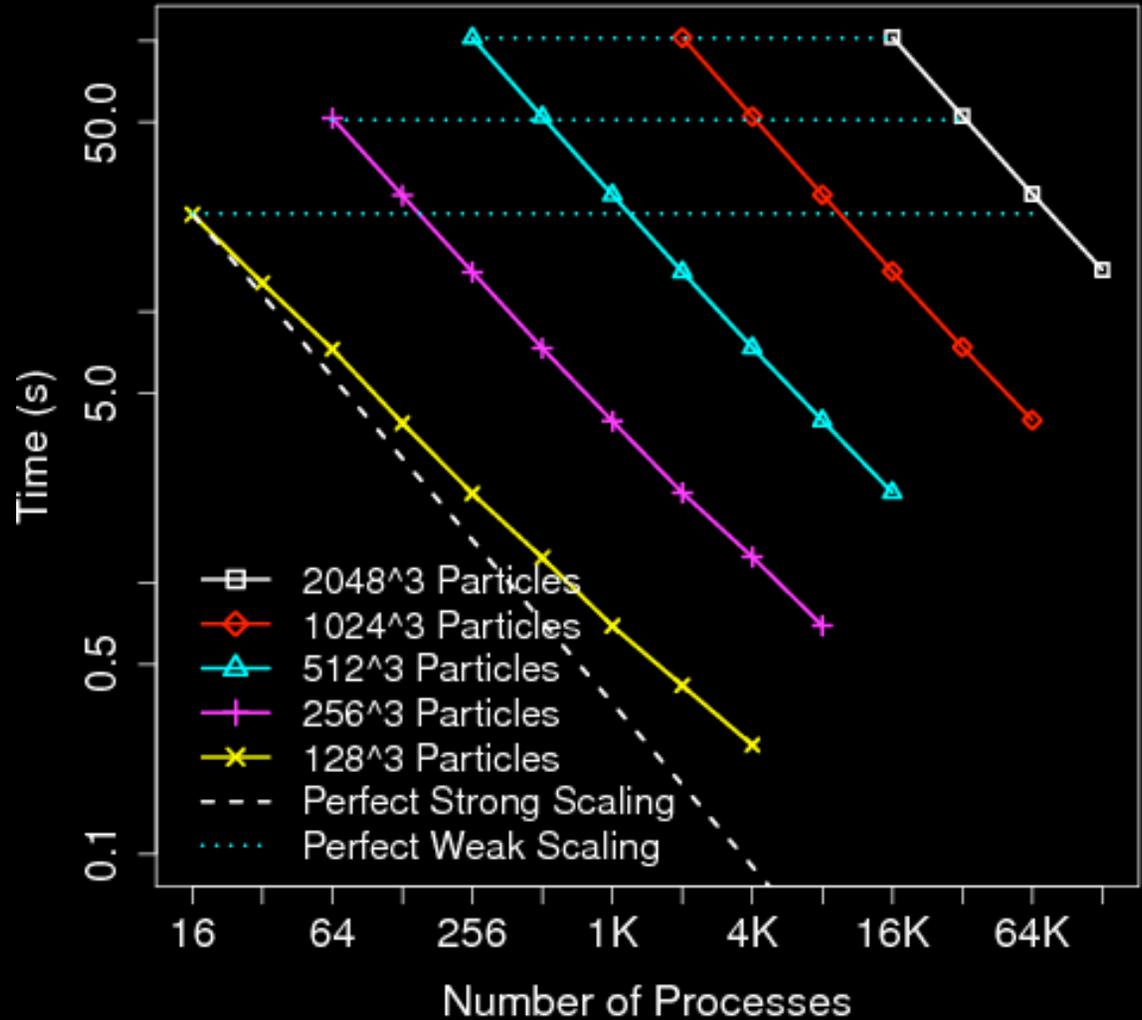
Computational Geometry

With Dmitriy Morozov and Carolyn Phillips

Strong and Weak Scaling with CGAL

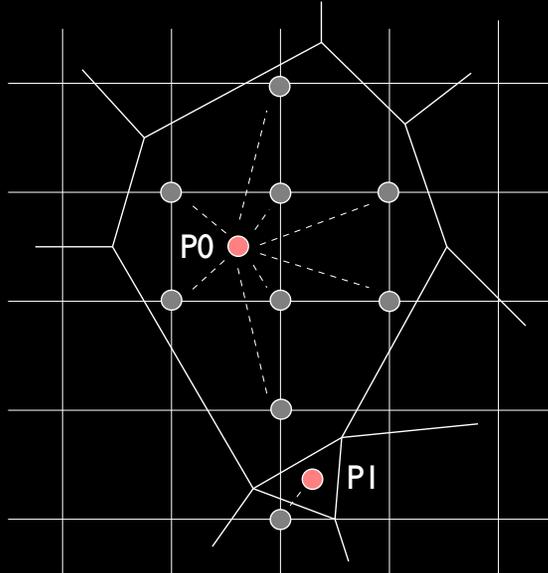


Strong and weak scaling for up to 2048^3 synthetic particles and up to 128K processes (excluding I/O) shows up to 90% strong scaling and up to 98% weak scaling.

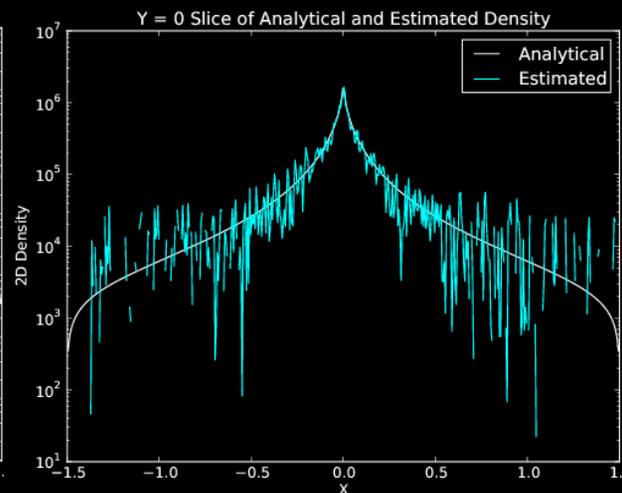
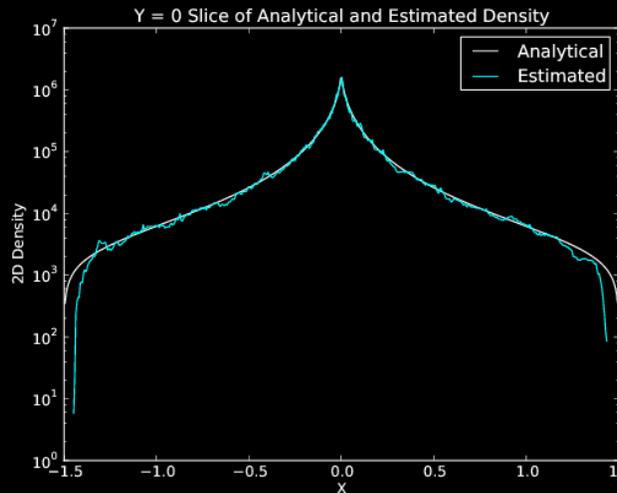
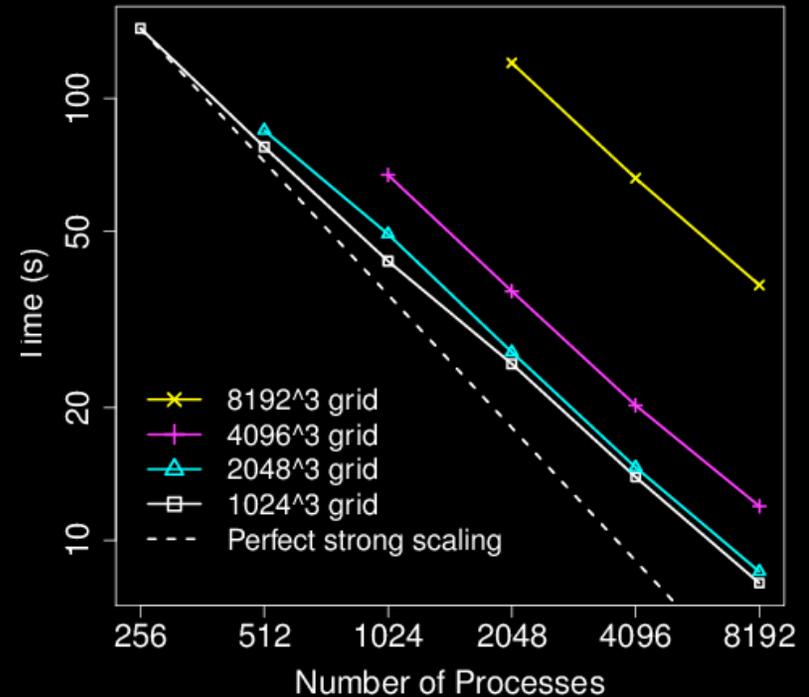


Tessellation-Based Density Estimation

Tessellation-based density estimation is parameter free, shape free, and automatically adaptive



Strong Scaling



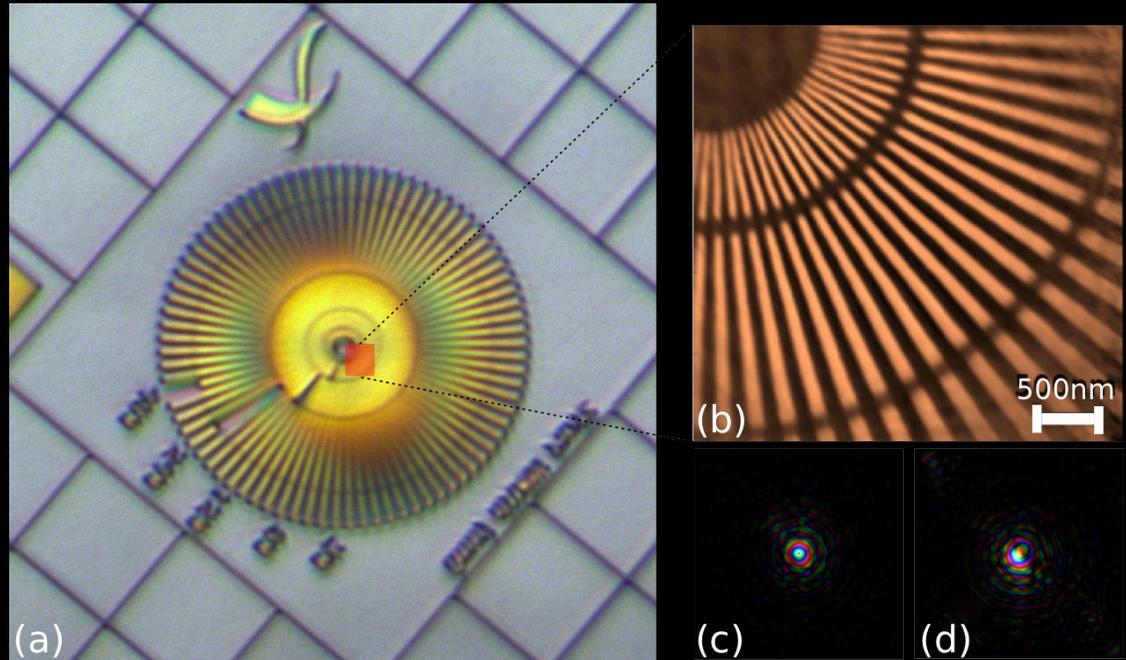
Above: Strong scaling of estimating the density of 512^3 synthetic particles onto grids of various sizes.

Left: comparison of tessellation-based and CIC density

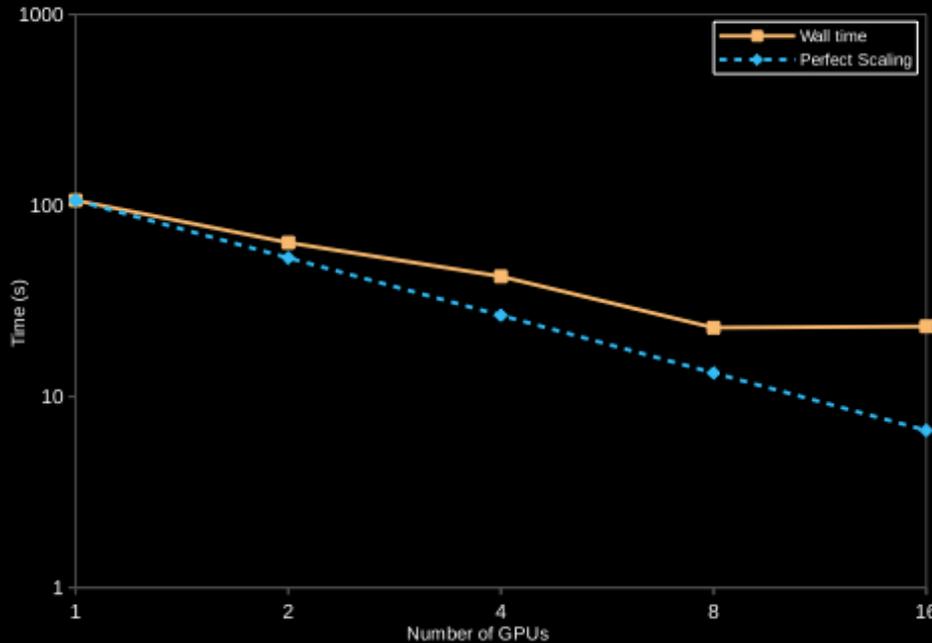
Parallel Reconstruction Performance

[Courtesy of Youssef Nashed, ANL]

A gold Siemens star test pattern, with 30 nm smallest feature size, was raster scanned through a 26×26 grid using a step size of 40nm and an exposure time of 0.6s per scan point, using a 5.2 keV X-ray beam. The total scanning time was about 20 minutes.



Strong Scaling on Real Data



Performance Breakdown on Real Data

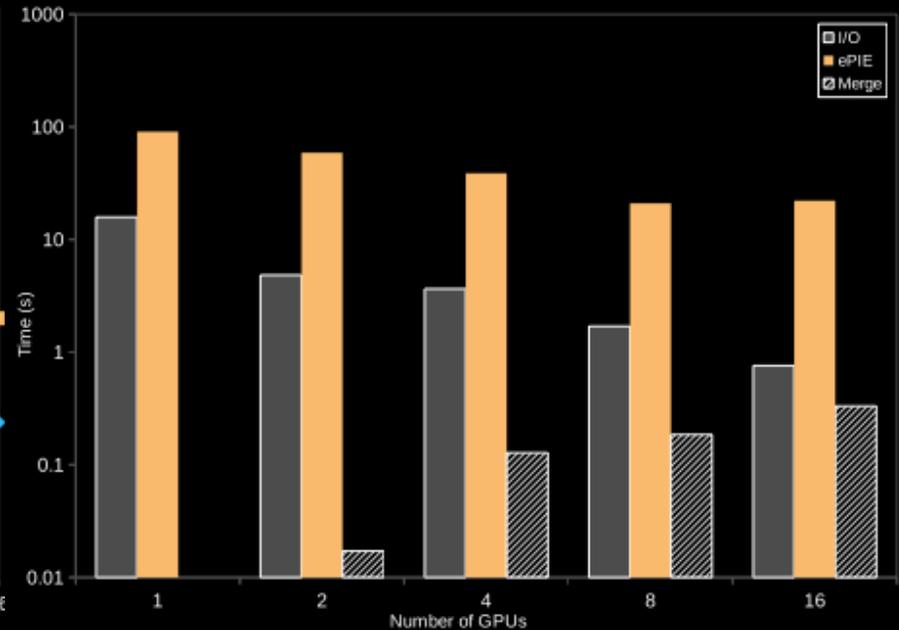
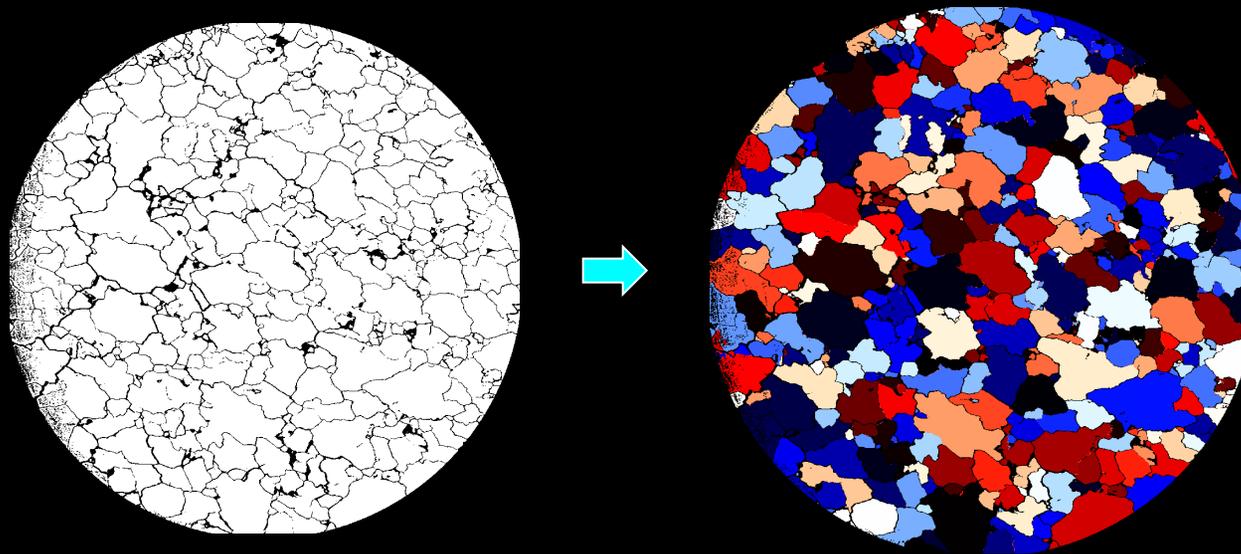


Image Segmentation in Porous Media

Courtesy Dmitriy Morozov and Patrick O'Neil



LBL (Dmitriy Morozov and Patrick O'Neil) developed tools for segmentation and connectivity analysis of granular and porous media using diy2.



Left: 3D image of a granular material (flexible sandstone) acquired at ALS by Michael Manga and Dula Parkinson. (Data: $2560 \times 2560 \times 1276$). Right: Watershed segmentation of the material identifies individual grains (run on Edison @ NERSC) [courtesy Morozov, O'Neil (LBL)].

Further Reading

DIY

- Peterka, T., Ross, R., Kendall, W., Gyulassy, A., Pascucci, V., Shen, H.-W., Lee, T.-Y., Chaudhuri, A.: Scalable Parallel Building Blocks for Custom Data Analysis. Proceedings of Large Data Analysis and Visualization Symposium (LDAV'11), IEEE Visualization Conference, Providence RI, 2011.
- Peterka, T., Ross, R.: Versatile Communication Algorithms for Data Analysis. 2012 EuroMPI Special Session on Improving MPI User and Developer Interaction IMUDI'12, Vienna, AT.

DIY applications

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- Lu, K., Shen, H.-W., Peterka, T.: Scalable Computation of Stream Surfaces on Large Scale Vector Fields. Proceedings of SCI4, New Orleans, LA, 2014.
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- Sewell, C., Meredith, J., Moreland, K., Peterka, T., DeMarle, D., Lo, Li-ta, Ahrens, J., Maynard, R., Geveci, B.: The SDAV Software Frameworks for Visualization and Analysis on Next-Generation Multi-Core and Many-Core Architectures. Proceedings of the SCI2 Ultrascale Visualization Workshop, Salt Lake City, UT.
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- Nouanesengsy, B., Lee, T.-Y., Lu, K., Shen, H.-W., Peterka, T.: Parallel Particle Advection and FTLE Computation for Time-Varying Flow Fields. Proceedings of SCI2, Salt Lake, UT.
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- Peterka, T., Kendall, W., Goodell, D., Nouanesengsey, B., Shen, H.-W., Huang, J., Moreland, K., Thakur, R., Ross, R.: Performance of Communication Patterns for Extreme-Scale Analysis and Visualization. Journal of Physics: Conference Series SciDAC 2010, 2010.

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People

Dmitriy Morozov, Han-Wei Shen, Abon Chaudhuri, Kewei Lu, Teng-Yok Lee, Attila Gyulassy, Valerio Pascucci, Salman Habib, Boothanome Nouanesengsy, Jian Huang, Rob Ross

<https://bitbucket.org/diatomic/diy>

Tom Peterka

tpeterka@mcs.anl.gov

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