

from the data, devices and interaction Laboratory

Thursday, February 20, 2020

Michael E. Papka

Northern Illinois University



Lectures and discussions of current research and technical developments in computer science for beginning graduate research students. Topics will emphasize open problems and recent scientific advances. Content may vary to reflect research advances in areas such as data analytics, scientific computing, graphics and visualization.

Lectures and discussions of current research and technical developments in computer science for beginning graduate research students. Topics will emphasize open problems and recent scientific advances. Content may vary to reflect research advances in areas such as data analytics, scientific computing, graphics and visualization.

Lectures and discussions of current research and technical developments in computer science for beginning graduate research students. Topics will emphasize open problems and recent scientific advances. Content may vary to reflect research advances in areas such as data analytics, scientific computing, graphics and visualization.

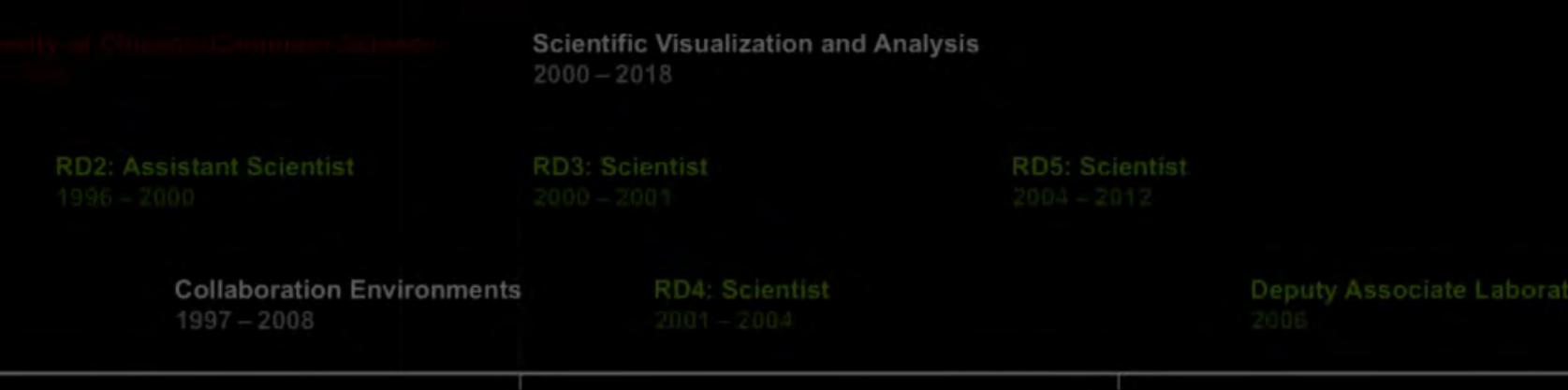
- Who has active research projects?
- What are NIU CS faculty interests?
- Where do I get more information?
- How do I get involved?

Bit About Me (Education)

- Northern Illinois University Physics (BS)
- University of Illinois @ Chicago Computer Science (MS)
- University of Chicago Computer Science (MS, PhD)
- Not the standard education ...
 - University of Chicago Business School (SLLP)
 - Harvard University Business School (GMP)
 - Stanford University Hasso Plattner Institute of Design (Design Thinking)

Bit About Me (Professional)

- Fermi National Accelerator Laboratory (Undergraduate/Graduate)
- Argonne National Laboratory
- Northern Illinois University



Bit About Me (Research - Areas/Interests

- Advanced Display Environments
- Collaboration Technology
- High Performance Computing (Environments)
- Information Visualization
- Scientific Visualization and Analysis
- Augmented/Virtual Reality

Collaboration Environments

Deputy Associate Laborat

RD5: Scientist

1997 - 2008

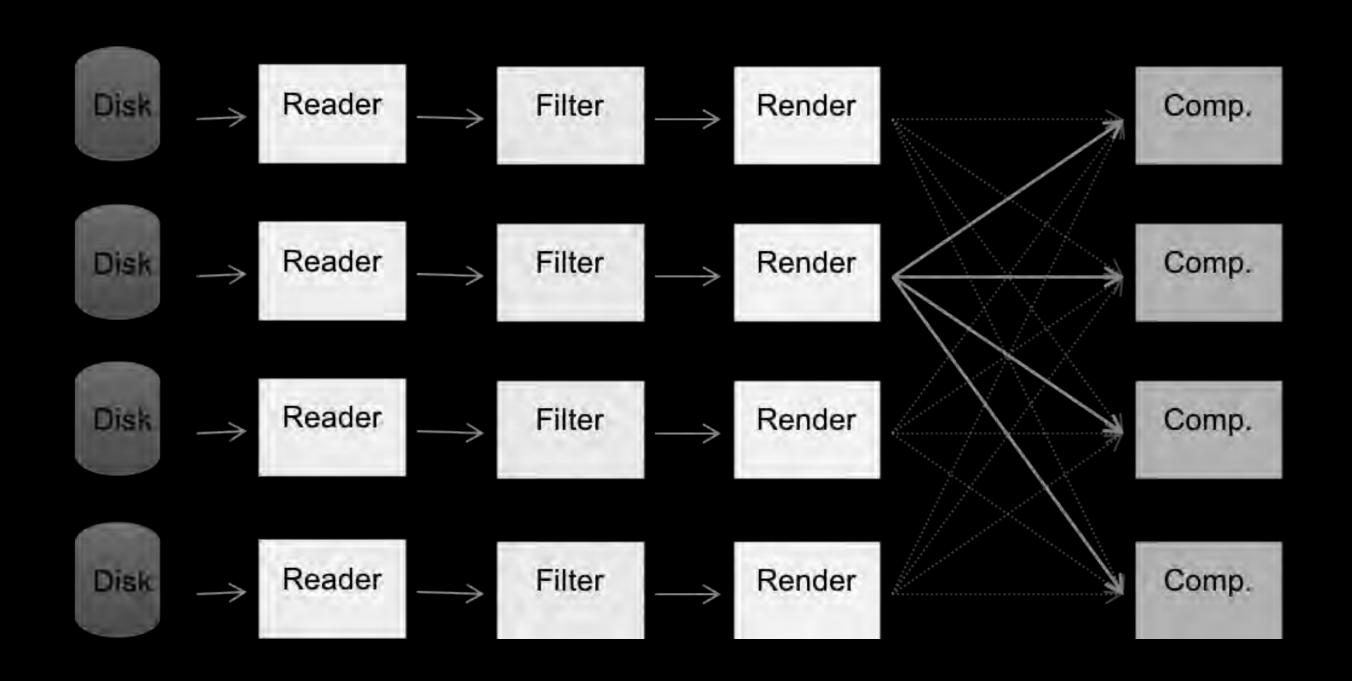
Bit About the ddiLab

- Joint lab with School of Art and Design Professor Joseph Insley from Time Arts
- Emphasis on visualization and data analysis coupled to highperformance computing in the support of science
- Students
 - 1 PhD (Information Visualization)
 - 3 MS (HPC log analysis, authentication infrastructure and machine learning)
 - 3 Undergraduates (virtual reality and HPC log analysis)

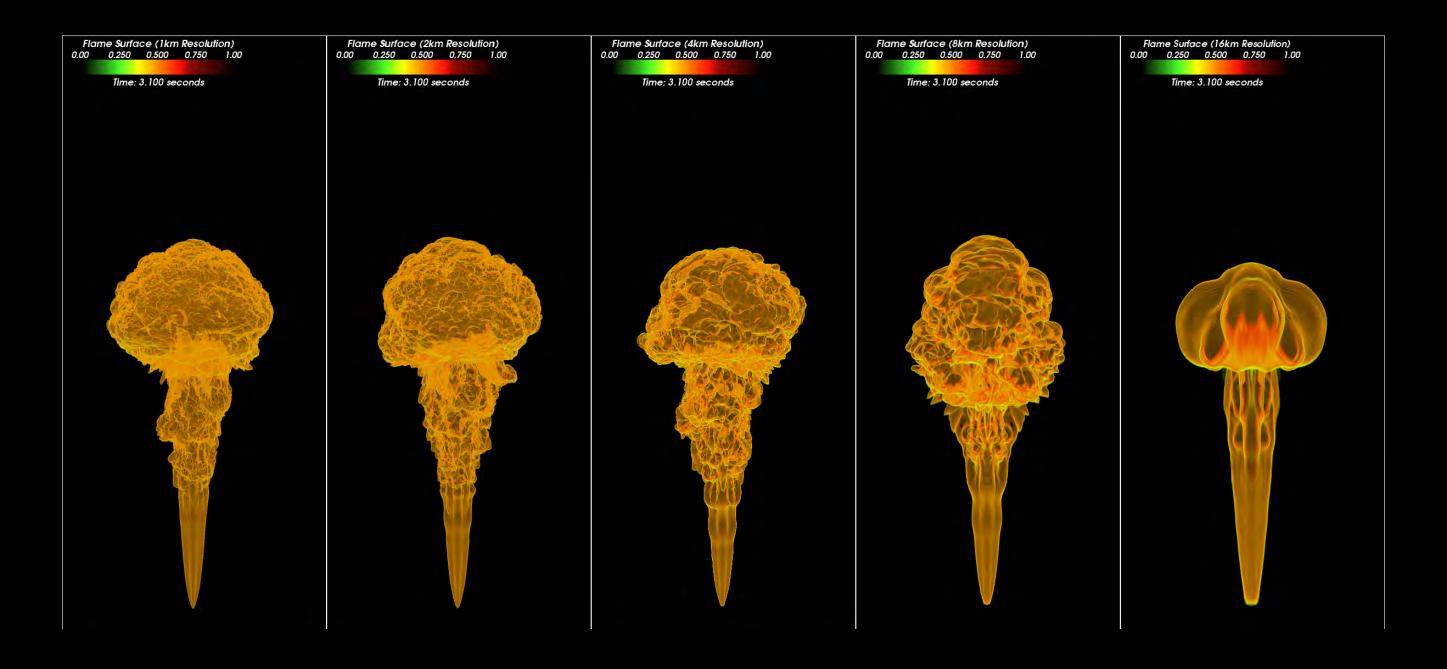
Scientific Visualization and Analysis

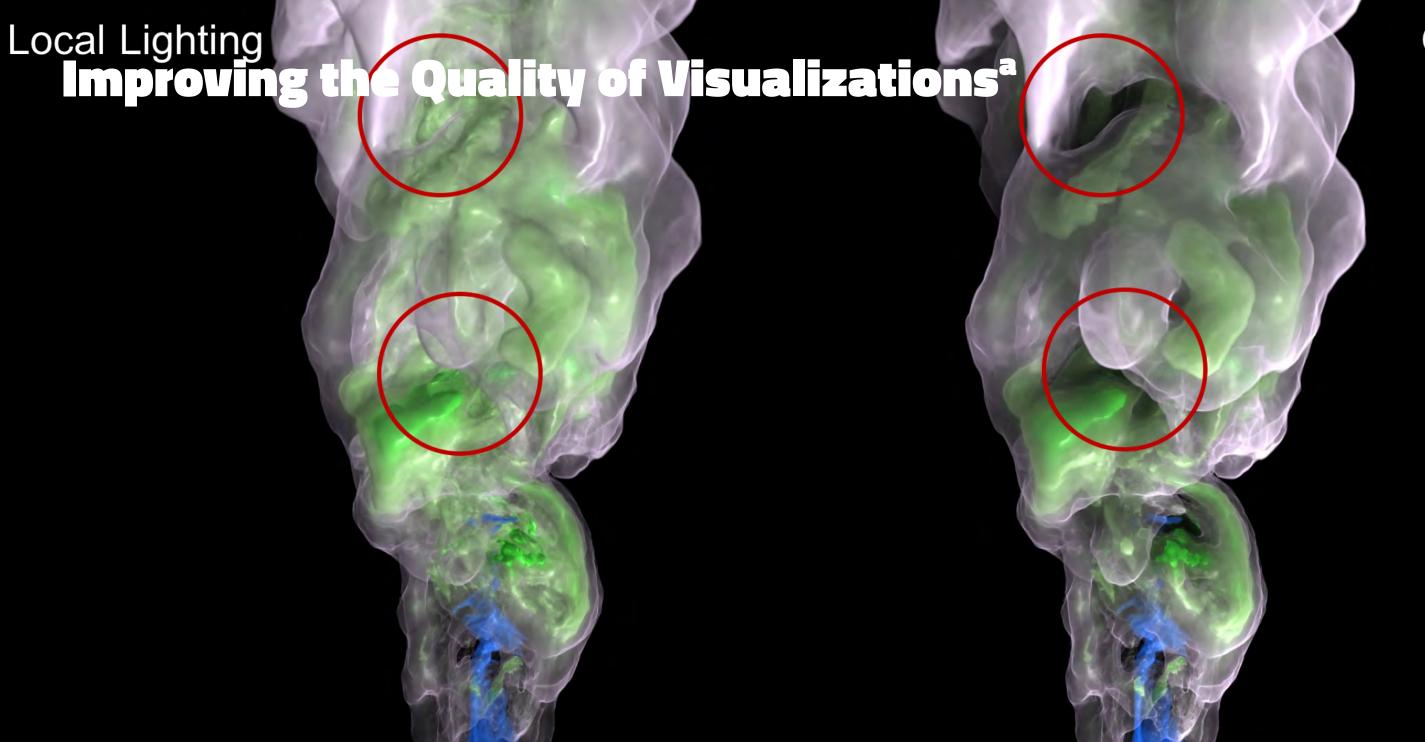
- vl3: volume rendering library
 - Parallel volume rendering library that exploits GPU hardware
 - Uses native data formats
- Integration with virtual and augmented reality
- Usability and collaboration
- Domain specific visualizations

vl3: Volume Rendering Library



Thermonuclear Flame Plume Rising in a Column

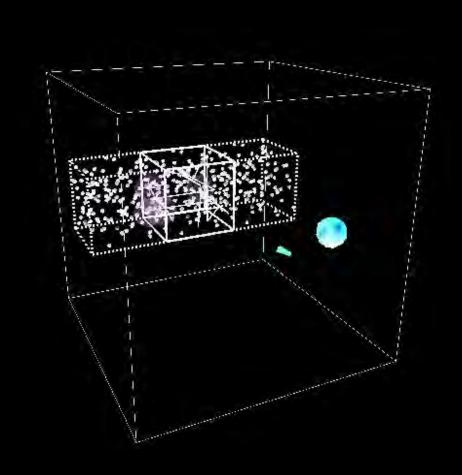




Global Lighting

^a M. Shih, S. Rizzi, J. Insley, T. Uram, V. Vishwanath, M. Hereld, M. E. Papka, K. L. Ma, *Parallel Distributed, GPU-Accelerated, Advanced Lighting Calculations For Large-Scale Volume Visualization*, 2016 IEEE 6th Annual Symposium on Large Data Analysis and Visualization (LDAV), pp. 47-55, October 2016.]

Virtual Reality bc

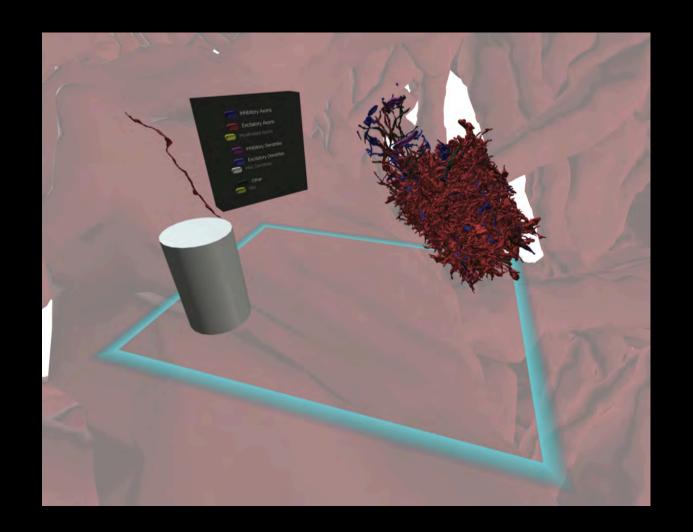




^bT. Disz, M. E. Papka, R. Stevens, M. Pellegrino, V. Taylor, *Virtual Reality Visualization of Parallel Molecular Dynamics Simulation*, **1995 Simulation Multiconference Symposium**, pp. 483-87, Phoenix, AZ, April 1995.

^c K. Reda, A. Knoll, K. Nomura, M. E. Papka, A. E. Johnson, J. Leigh, *Visualizing Large-Scale Atomistic Simulations in Ultra-resolution Immersive Environments*, **Proceedings of the 2013 IEEE Symposium on Large Data Analysis and Visualization (LDAV 2013)**, pp. 59-66, Atlanta, GA, October 13-14, 2013.

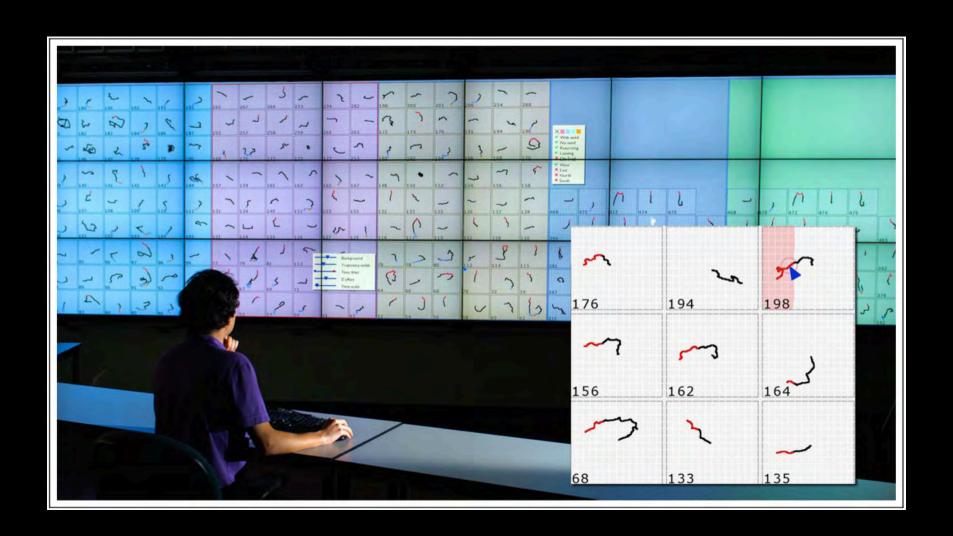
Virtual Reality^d





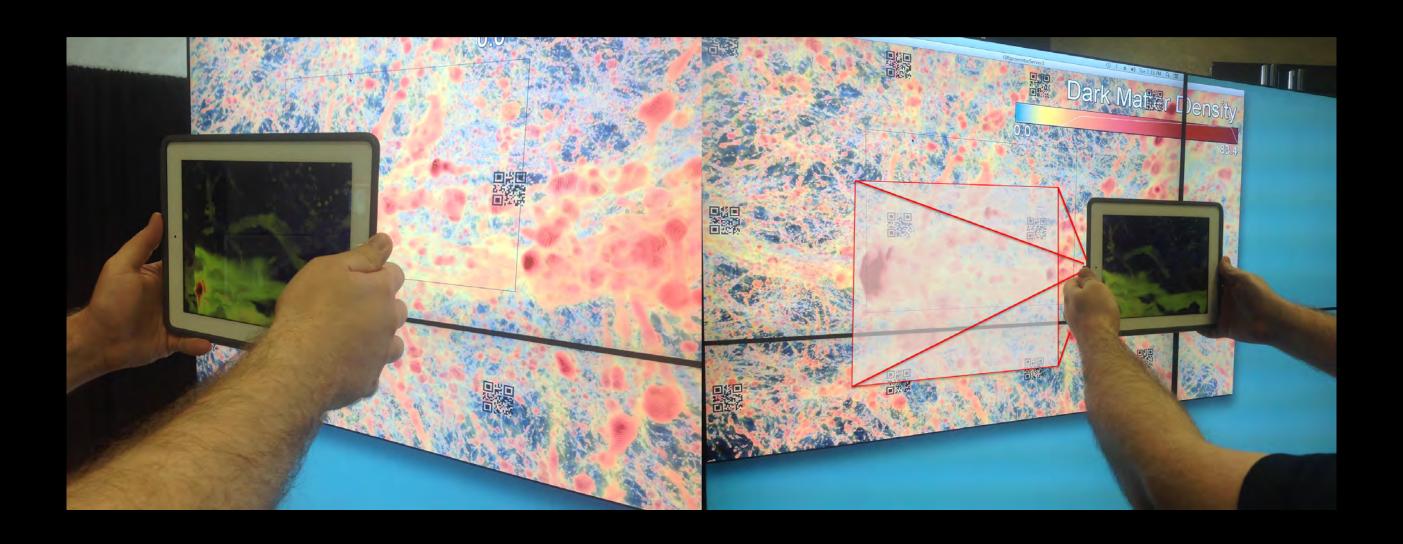
^d E. B. Brooks, J. A. Insley, M. E. Papka, S. Rizzi, *Virtual reality tools for the correction of automated volume segmentation errors using dense surface reconstructions*, **2017 IEEE 7th Symposium on Large Data Analysis and Visualization (LDAV)**, pp. 92-93, October 2, 2017. [POSTER]

Usability and Collaboration^e



^e K. Reda, A. E. Johnson, M. E. Papka, J. Leigh, *Modeling and Evaluating User Behavior in Exploratory Visual Analysis*, **Information Visualization** 15(4), pp. 325-339, October 2016.

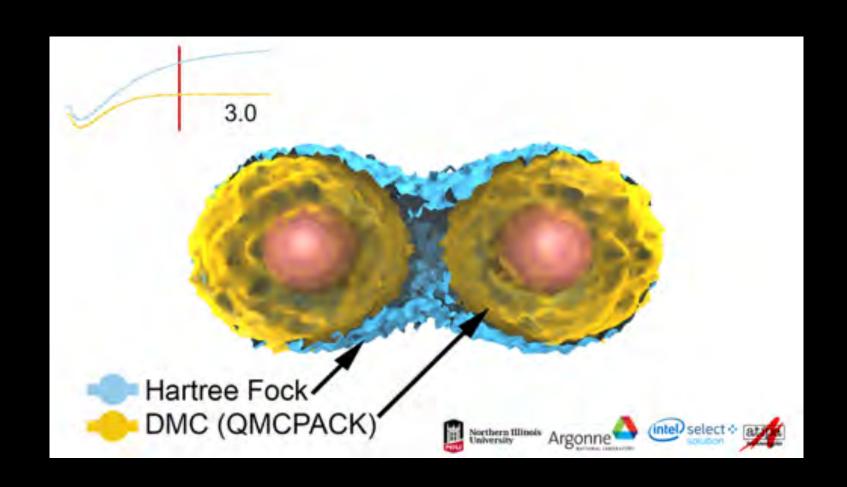
Usability and Collaboration^f



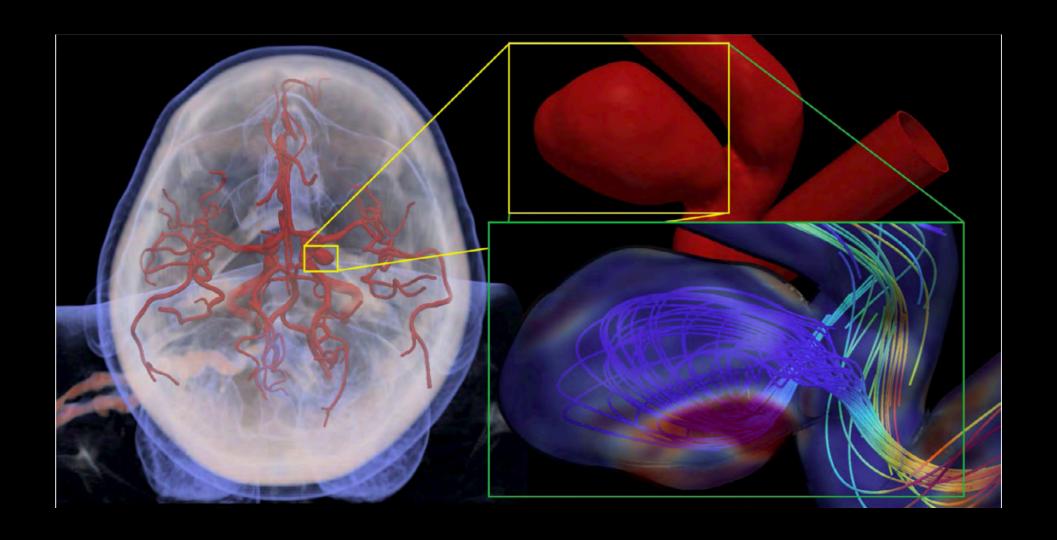
^fP. Lindner, A. Rodriguez, T. Uram, M. E. Papka, *Augmenting Views on Large Format Displays with Tablets*, Proceedings of the 2nd ACM Symposium on Spatial User Interaction (SUI 2014), Honolulu, HI, October 4-5, 2014. [Poster]

Domain Specific Visualizations

- Applied solutions to specific problems within domain
- Deep partnership with domain experts
- Current effort with NIU Chemistry
 - Visualizing and Quantifying Structural Ordering Underlying Static Structure Factor Peaks from Molecular Dynamics Simulations Travis Mackoy, Bharat Kale, Ralph Wheeler
 - Comparison Visualizations of Electron Density Approximation Methods Anouar Benali (ANL), Joe Insley, Ralph Wheeler

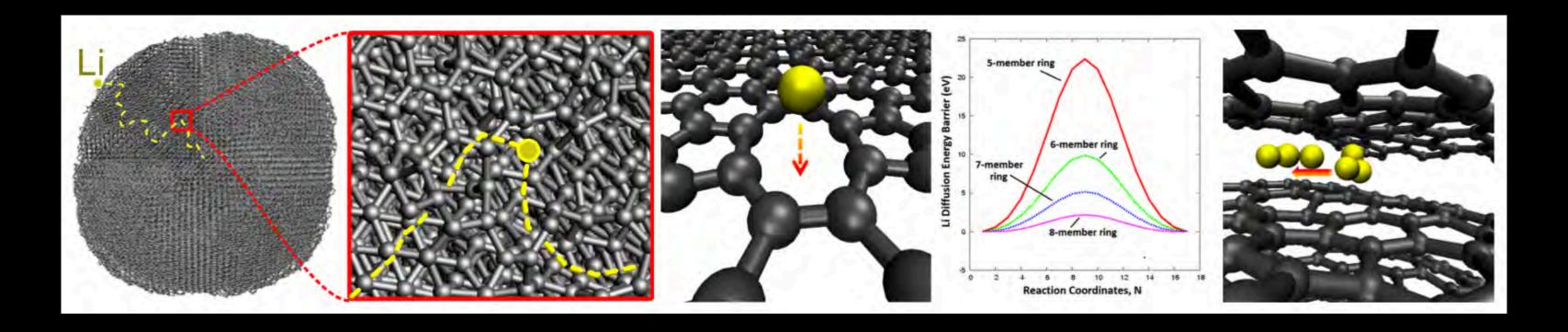


Domain Specific Visualizations^g



^g P. Perdikaris, J.A. Insley, L. Grinberg, Y. Yu, M. E. Papka, G. E. Karniadakis, *Visualizing Multiphysics, Fluid-Structure Interaction Phenomena in Intracranial Aneurysms*, Parallel Computing, 55, pp. 9-16, July 2016.

Domain Specific Visualizations^h



^h A. Gyulassy, A. Knoll, K. C. Lau, B. Wang, P.-T. Bremer, M. E. Papka, L. Curtiss, V. Pascucci, *Interstitial and Interlayer Ion Diffusion Geometry Extraction in Graphitic Nanosphere Battery Materials*, **IEEE Transactions on Visualization and Computer Graphics**, 22(1):916-925, January 2016.

Domain Specific Visualizations



Domain Specific Visualizations

High Performance Computing

- Applicationsⁱ
- Communication^j
- Operations^k

R. Fisher, L. Kadanoff, D. Lamb, A. Dubey, T. Plewa, A. Calder, F. Cattaneo, P. Constantin, I. Foster, M. E. Papka, S. I. Abarzhi, S. M. Asida, P. M. Rich, C. C. Glendenin, K. Antypas, D. J. Sheeler, L. B. Reid B. Gallagher, and S. G. Needham, *Terascale Turbulence Computation Using the FLASH3 Application Framework on the IBM Blue Gene/L System*, IBM Journal of Research and Development, 52(1.2):127-36, 2008.

¹V. Vishwanath, M. Hereld, V. Morozov, M. E. Papka, *Topology-Aware Data Movement and Staging for I/O Acceleration on Blue Gene/P Supercomputing Systems*, **SC'11 Proceedings of 2011 International Conference for High Performance Computing, Networking, Storage and Analysis**, Article No. 19, Seattle, WA, November 2011.

*S. Read, M. E. Papka, Operational Metrics Reporting Processes at Scientific User Facilities: Comparing A High-Energy X-Ray Synchrotron Facility to a Supercomputing Facility}, 2017 IEEE International Professional Communication Conference (ProComm), pp. 1-6, Madison, WI, July 23, 2017.

High Performance Computing

- Power
- Scheduling^m
- Workflows/Workloads^{n,o}

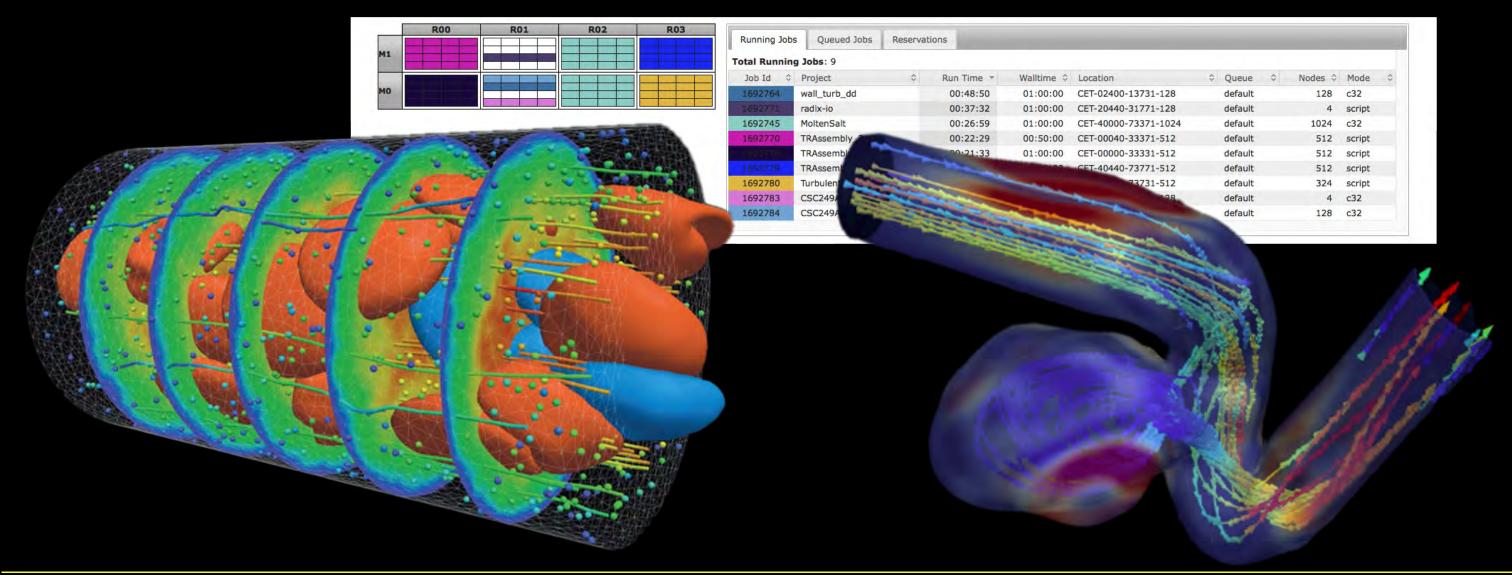
S. Wallace, Z. Zhou, V. Vishwanath, S. Coghlan, J. Tramm, Z. Lan, M. E. Papka, *Application Power Profiling on IBM Blue Gene/Q*, Parallel Computing, 57, pp. 73-86, September 2016.

^m Y. Fan, Z. Lan, P. Rich, W. E. Allcock, M. E. Papka, B. Austin, D. Paul, *Scheduling Beyond CPUs for HPC*, **Proceedings of the 28th International Symposium on High-Performance Parallel and Distributed Computing**, pp. 97-108, June 2019.

W. E. Allcock, B. S. Allen, R. Ananthakrishnan, B. Blaiszik, K. Chard, R. Chard, I. Foster, L. Lacinski, M. E. Papka, R. Wagner, *Petrel: A Programmatically Accessible Research Data Service*, **Proceedings of the Practice and Experience in Advanced Research Computing on Rise of the Machines**, pp. 49, July 2019.

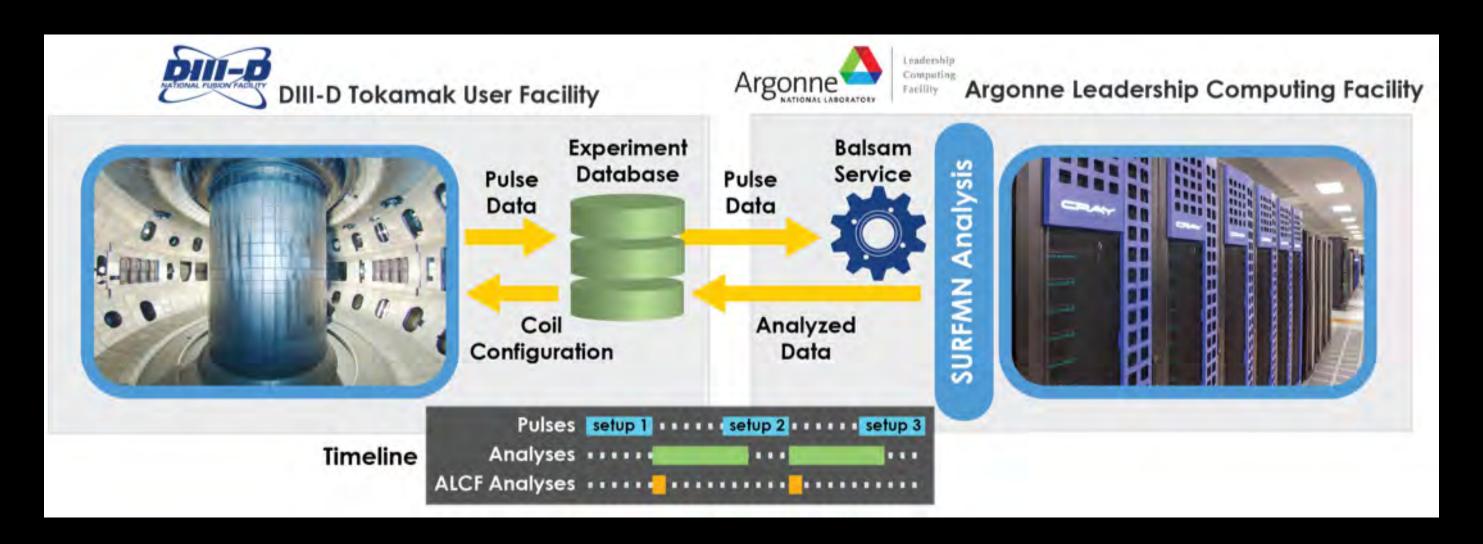
^o M. A. Salim, T. D. Uram, J. T. Childers, P. Balaprakash, V. Vishwanath, M. E. Papka, *Balsam: Automated Scheduling and Execution of Dynamic, Data-Intensive HPC Workflows*, arXiv preprint arXiv:1909.08704, September 2019.

Traditional^p



^pL. Grinberg, J. A. Insley, D. Fedosov, V. A. Morozov, M. E. Papka, G. E. Karniadakis, *Tightly Coupled Atomistic-Continuum Simulations of Brain Blood Flow on Petaflop Supercomputers*, **Computing in Science and Engineering**, 14(6):58-67, 2012.]

Evolving (scheduling constraints)^q

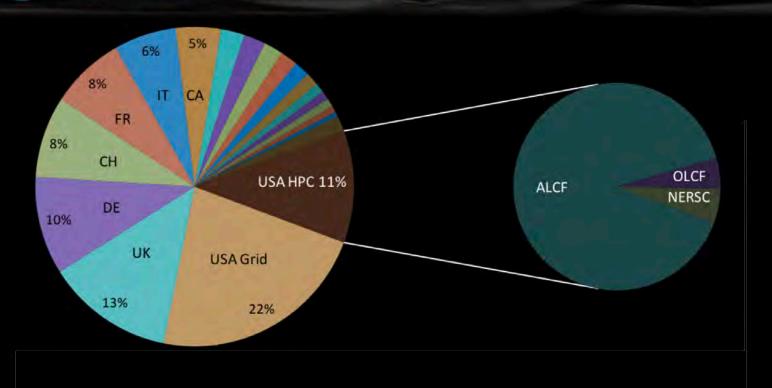


^qM. Kostuk, T. D. Uram, T. Evans, D. M. Orlov, M. E. Papka, D. Schissel, *Automatic Between-Pulse Analysis of DIII-D Experimental Data Performed Remotely on a Supercomputer at Argonne Leadership Computing Facility*, **Fusion Science and Technology**, September 2017.]

Evolving (complex workflows)^r

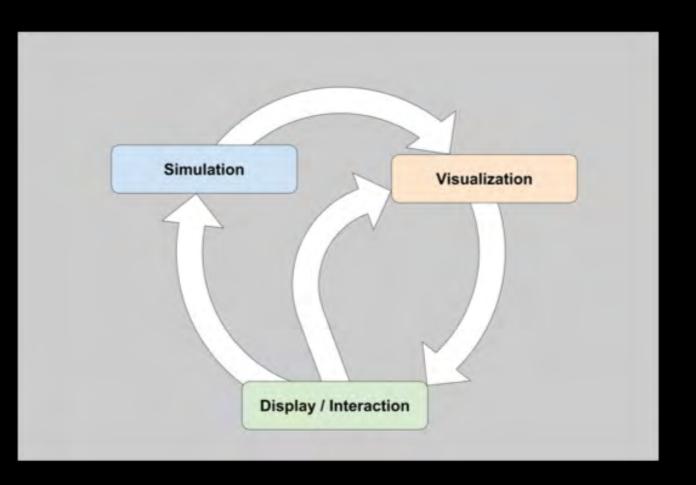
50% of the ATLAS papers based on 2015 data use the HPC-produced computing in a demonstrable manner

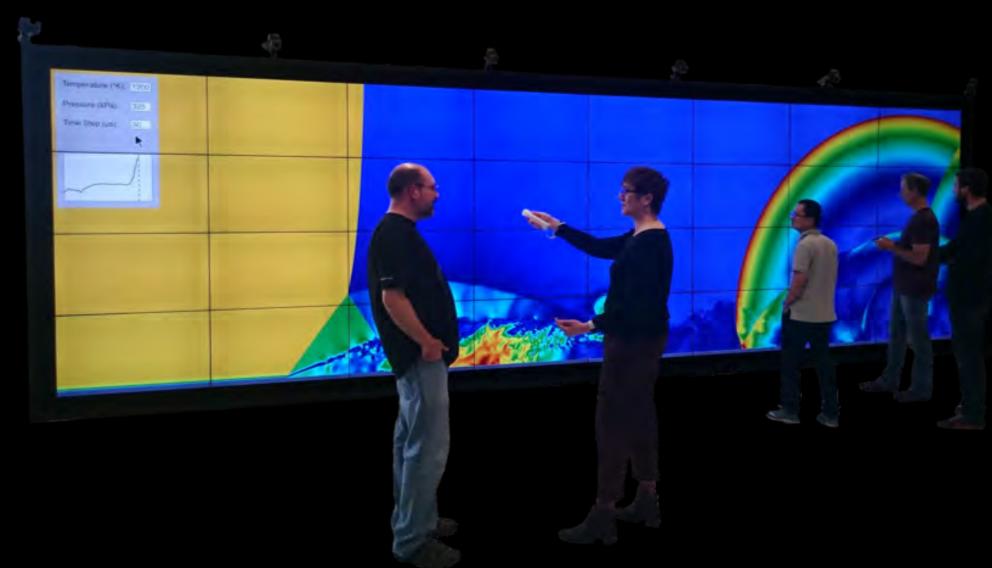
 These would still eventually be written without the US HPC effort, but they probably would not exist today: the time-to-science has been dramatically shortened.



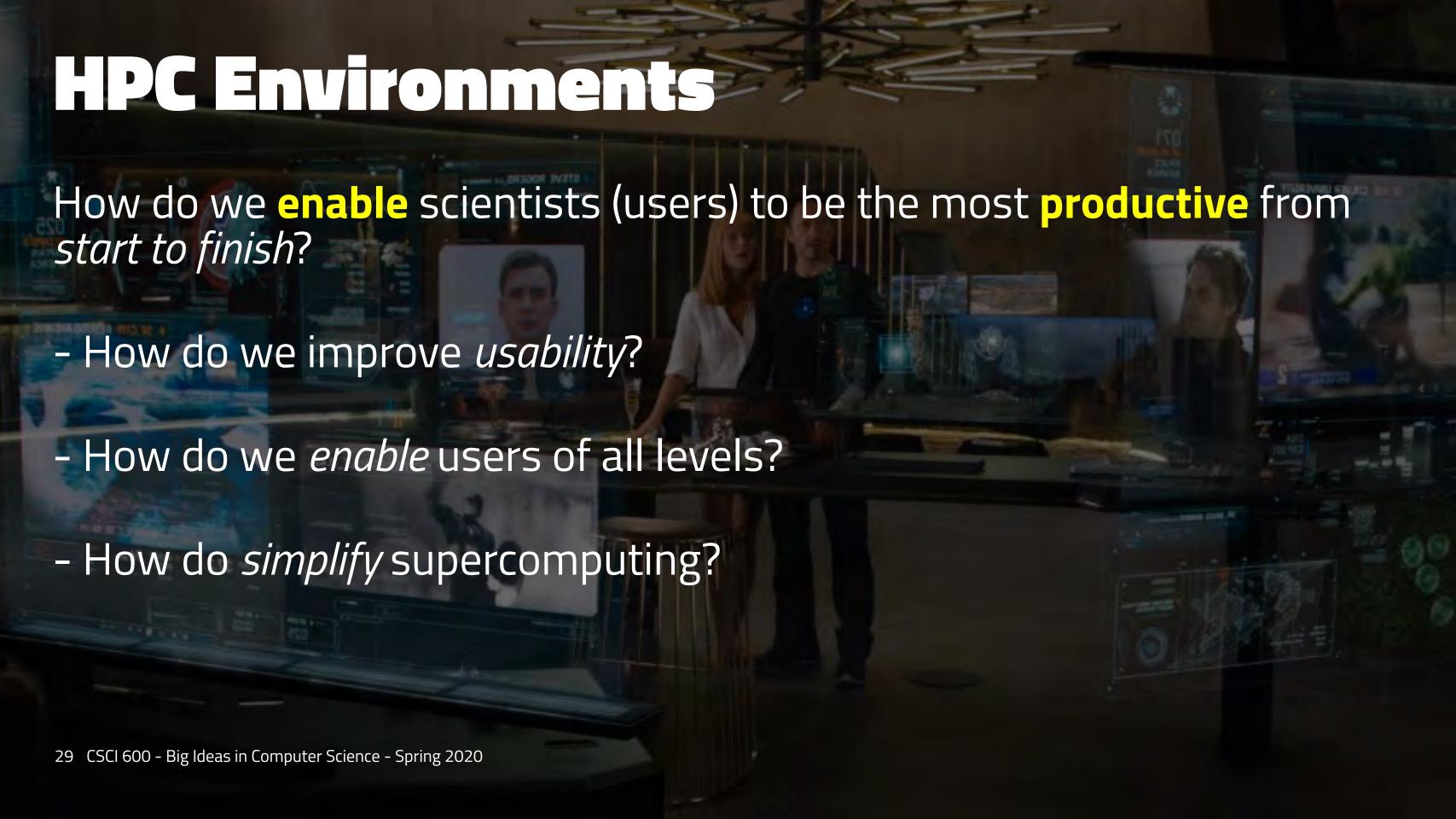
T. LeCompte(HEP){06/02/2016} and J. T. Childers, T. D. Uram, D. Benjamin, T. J. LeCompte, M. E. Papka, An Edge Service for Managing HPC Workflows, Proceedings of the Fourth International Workshop on HPC User Support Tools (HUST'17), Denver, CO, November 12, 2017.]

Evolving (increased engagement)^s

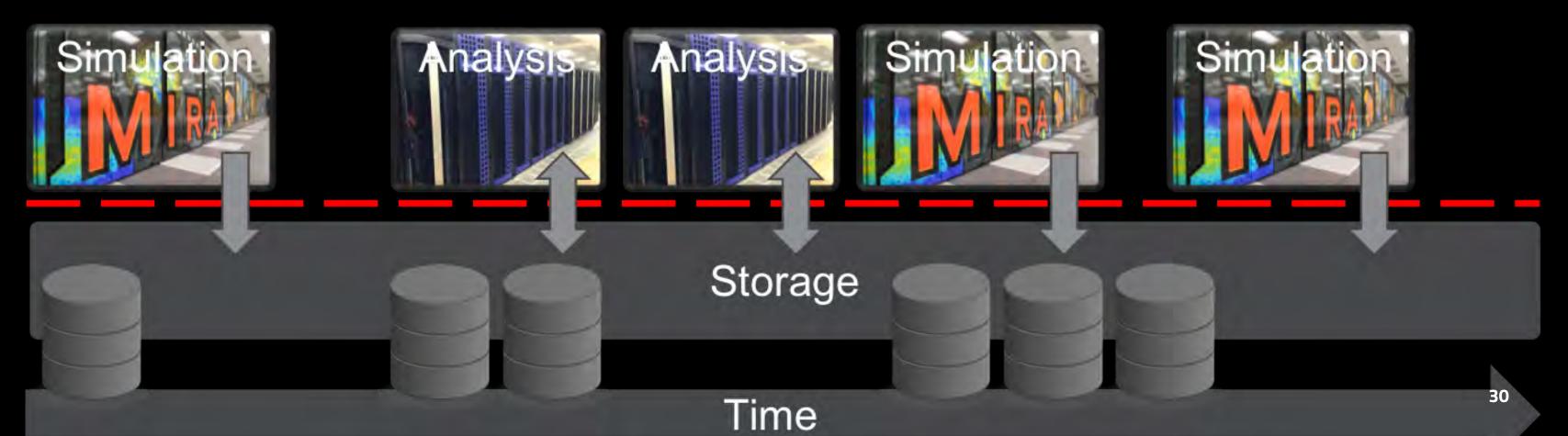




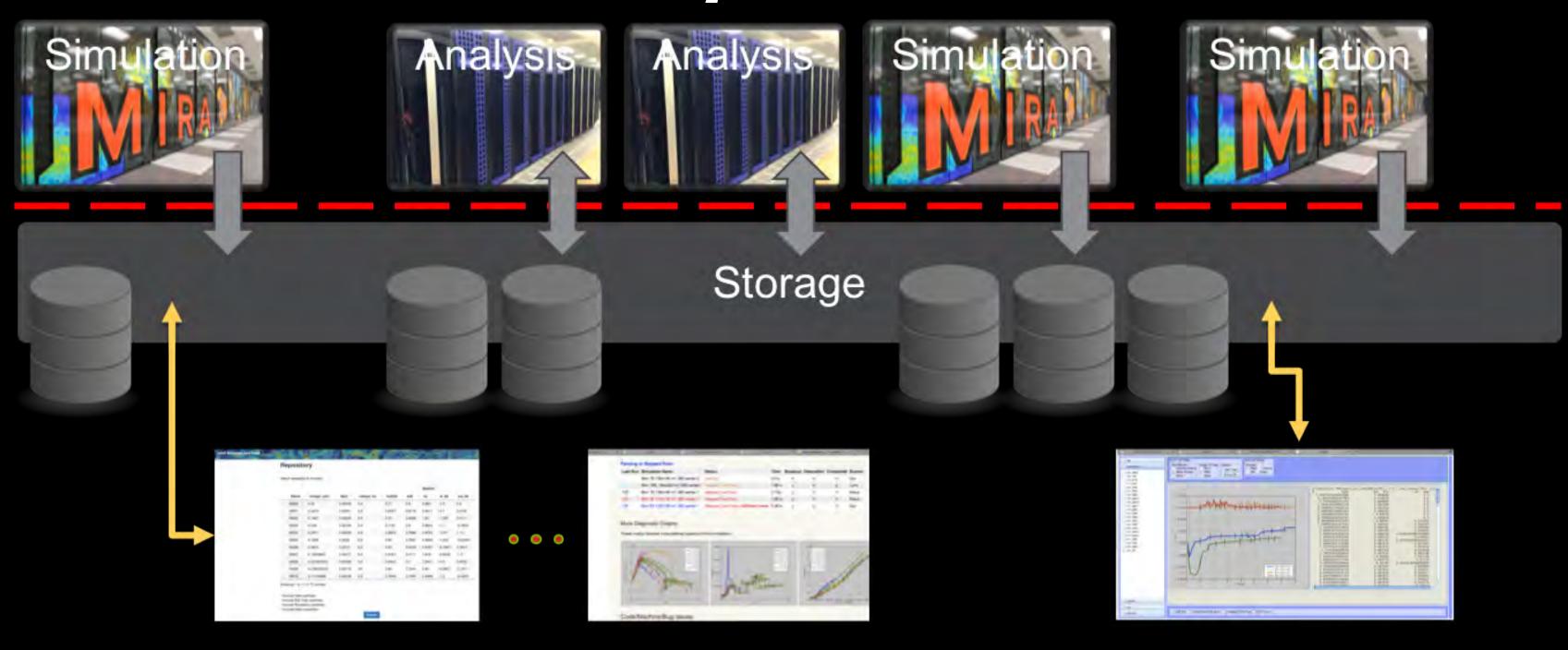
^sT. Marrinan, A. Nishimoto, J. A. Insley, S. Rizzi, A. Johnson, M. E. Papka, *Interactive Multi-Modal Display Spaces for Visual Analysis*, **Proceedings of the 2016 ACM on Interactive Surfaces and Spaces**, pp. 421-426, Niagara Falls, Canada, November 6, 2016.]



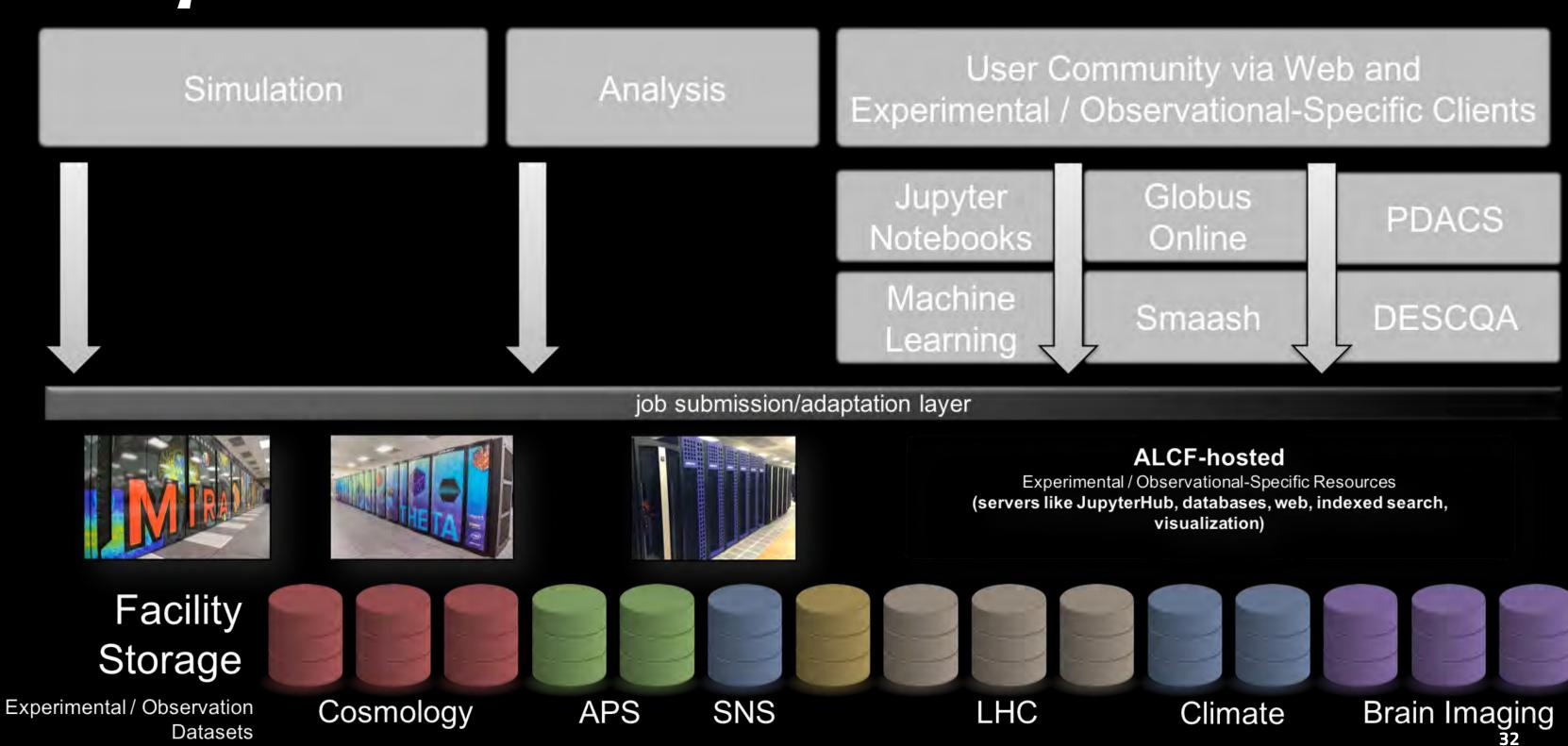
Workflow of Today

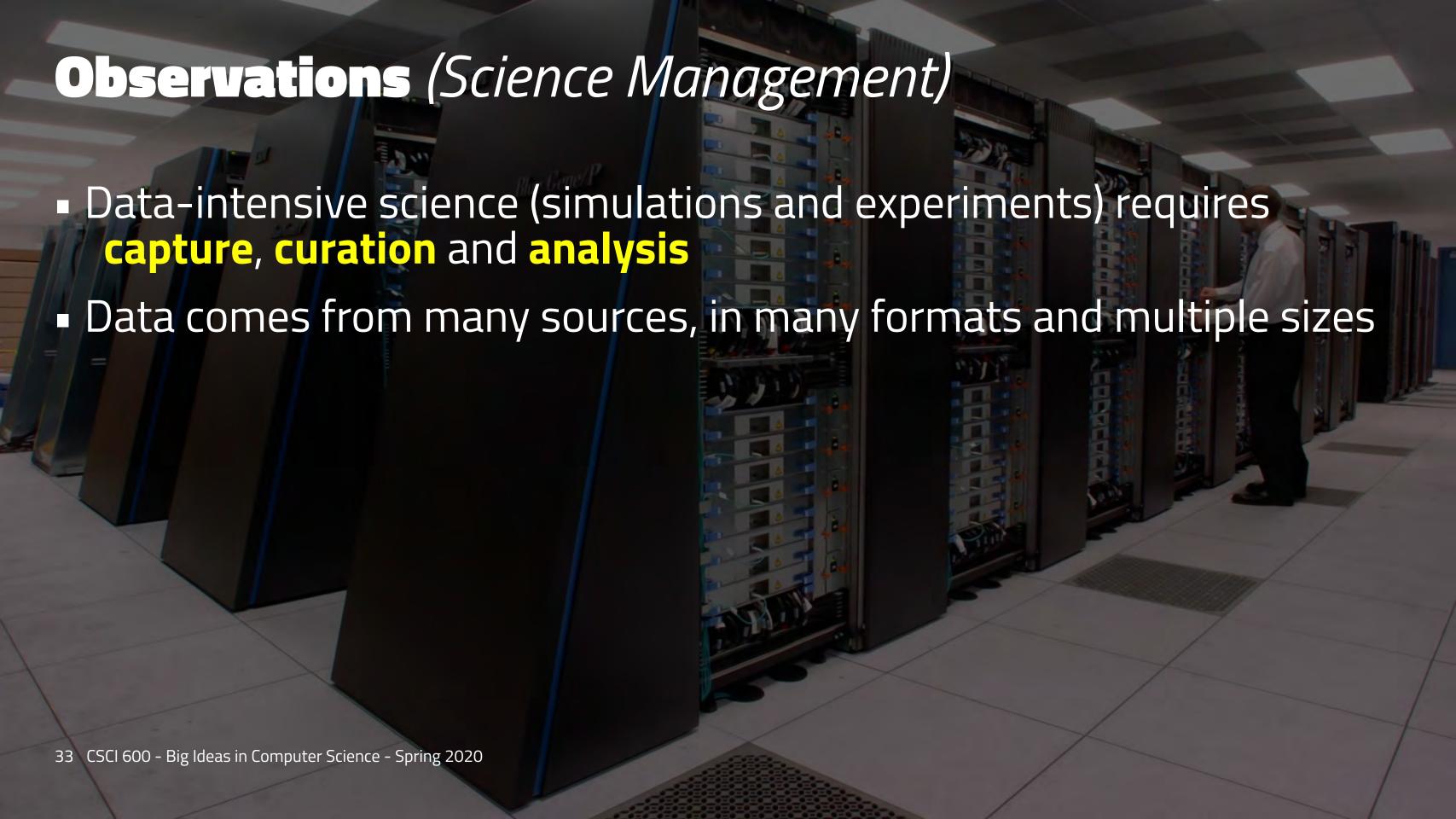


Workflow of Tomorrow (Today)



Facility of Tomorrow





Observations (Science Management)

- Problem with science management:
 - Tracking simulations and output [difficult]
 - Finding and reproducing old simulations: [difficult]
 - Monitoring live simulations: [inconvenient, idiosyncratic]
 - Post-processing, analysis and archival of results: [haphazard]
 - Assessing simulation behavior/performance: [difficult]

Increased Access to Scientific Communities

Support for Application Teams

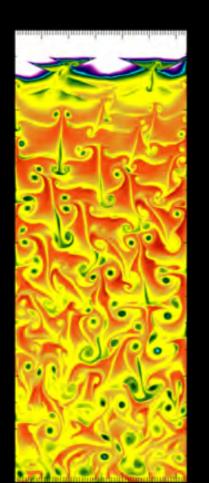
Simulation management and analysis system for Flash (Smaash)^t

- Tracking and coordination of data (simulation and meta)
- Run-time monitoring of simulations and automated analysis of simulation output
- Method for managing / executing common workflows

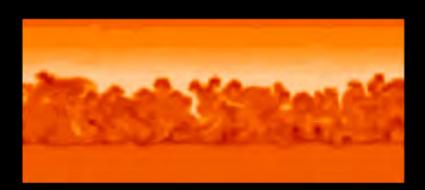
[†]R. Hudson, J. Norris, L. B. Reid, K. Weide, G. C. Jordan, and M. E. Papka, *Experiences Using Smaash to Manage Data-Intensive Simulations*, **Proceedings of the 20th International Symposium on High-Performance Parallel and Distributed Computing**, pp. 205-15, San Jose, CA, June 2011.

Prototype Partner - Flash

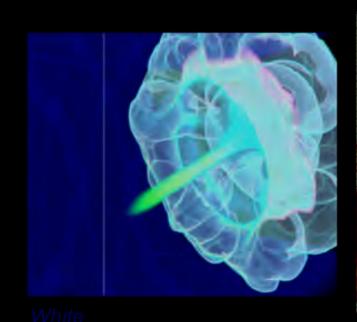
- Multi-physics
- Adaptive-mesh

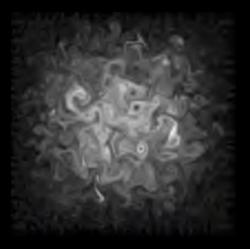


Cellular detonations

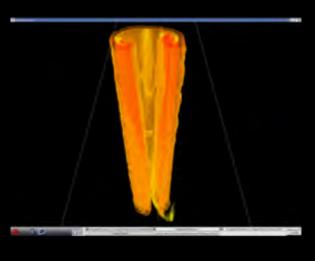


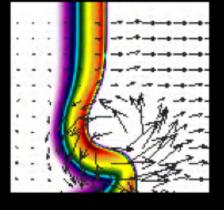
Nova outbursts on white dwarfs



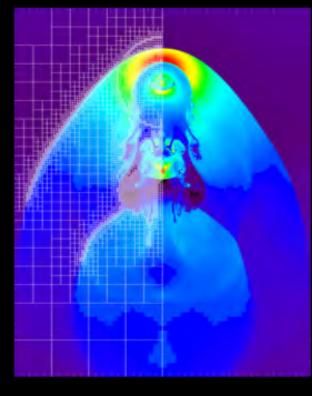


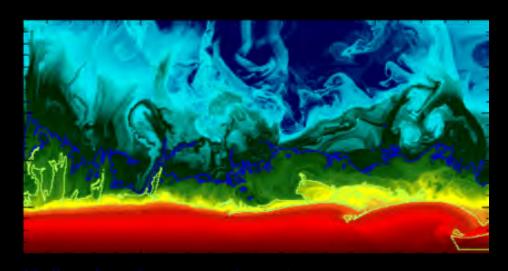
Compressible turbulence





Flame-vortex interactions





Helium burning on neutron stars

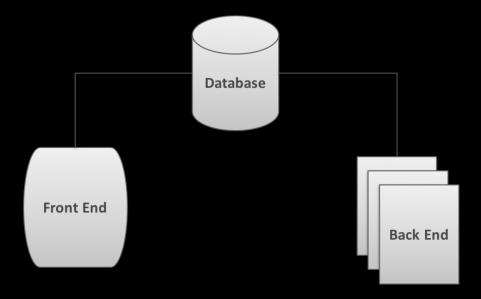
Ravleigh-Taylor instability

Prototype Partner - Flash

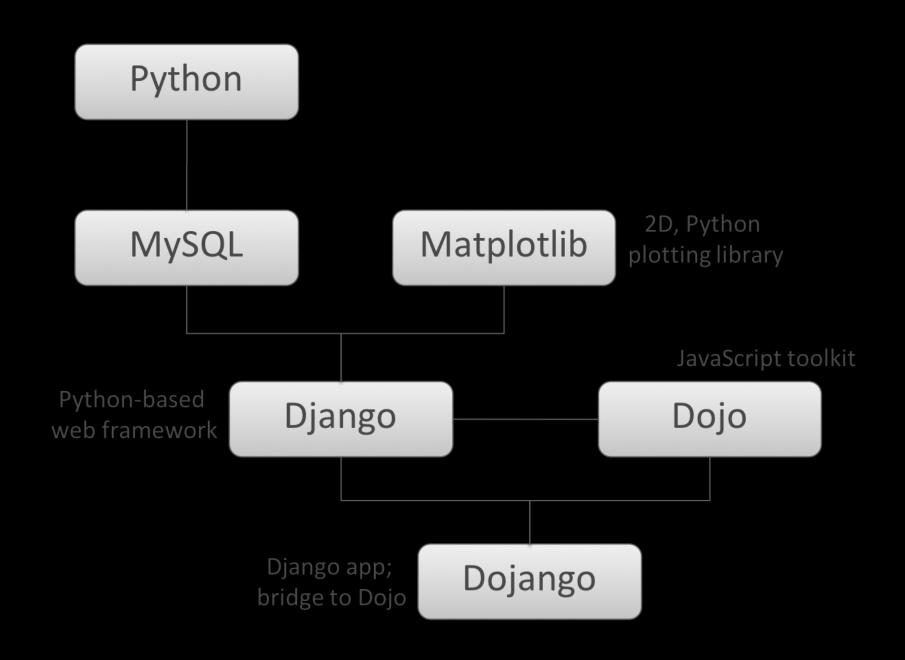
- Meta-data output
 - log: simulation progress, warnings, errors, resource use
 - .dat: integrated grid quantities
- Scientific data output (HDF5)
 - Checkpoint: complete information needed to restart simulation
 - Plotfile: data values of interest for analysis
 - Particle files: tracer particles of interest during analysis

Smaash Components

- Database (manages meta-data)
- Back end services (co-located with compute resources and scientific data)
- Front end interfaces (user facing)



Smaash Implementation



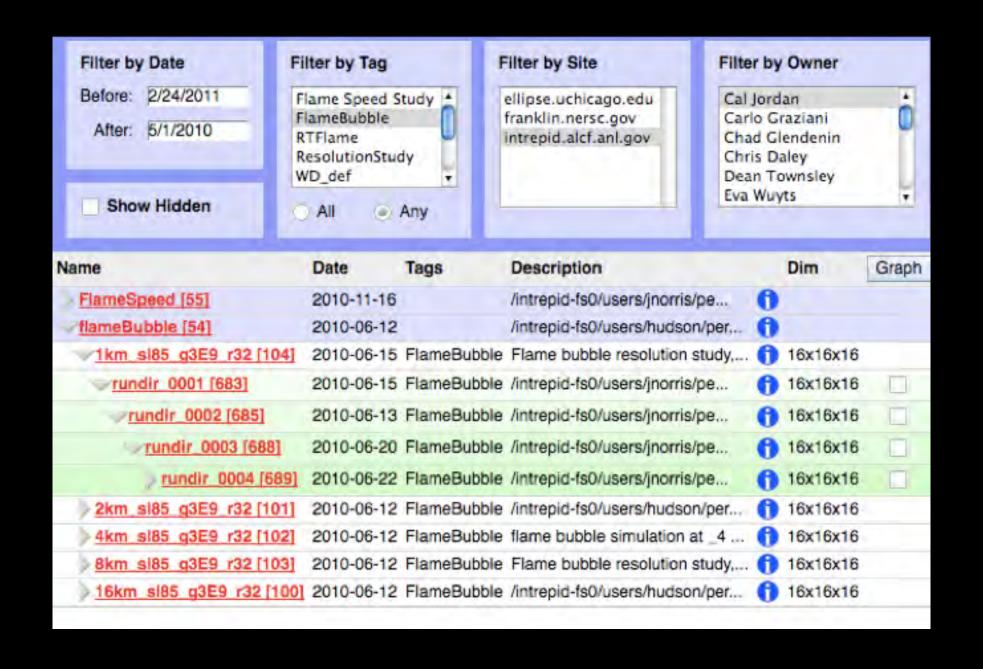
Smaash Back End Services

- Collector captures and stores meta-data in database about simulation
- Archiver automates the archiving of data
- Verifier cross checks output and database entries
- Associator connects a current simulation with campaign
- Observer responsible for updates to user (email)
- Visualizer automatic running of user specified visualization scripts

Smaash Front End Interfaces (Views)

- Tree collection of campaigns, simulations and runs
- Graph quick graphs of results
- Monitor automated visualizations
- Summary details and notes

Tree View

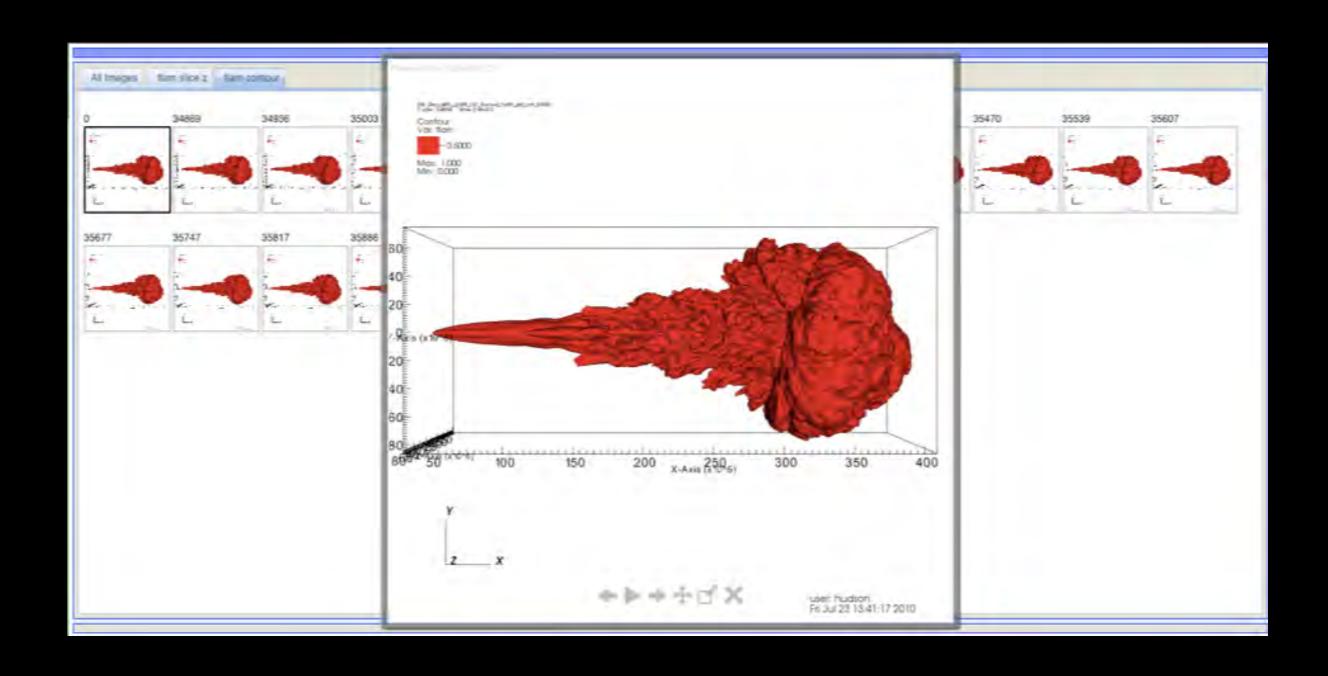


Graph View

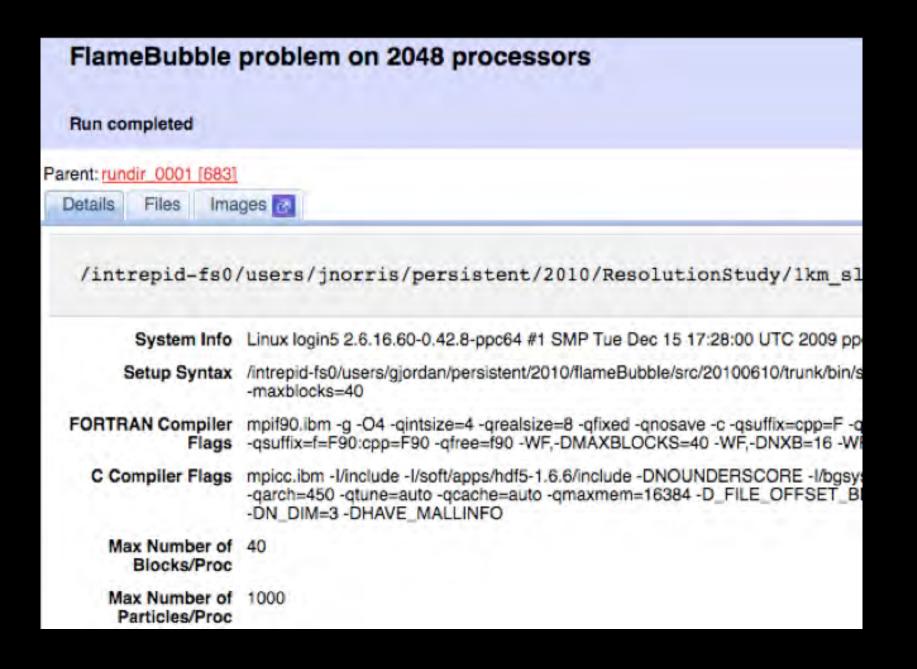


http://flashdb.ci.uchicago.edu/graphBranches/410,425/using/v90/vs/v32/cstroke/png

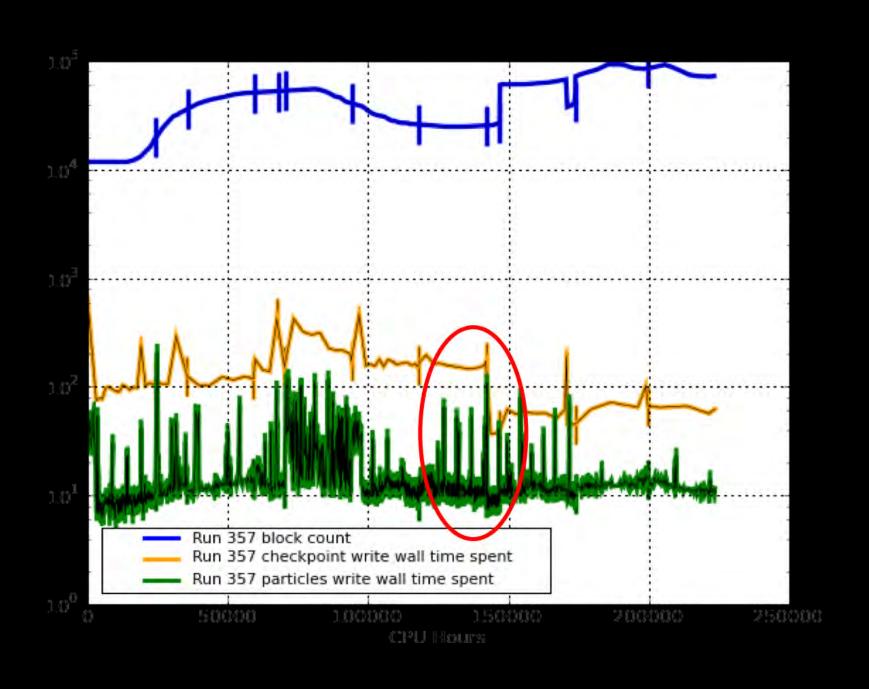
Monitor View



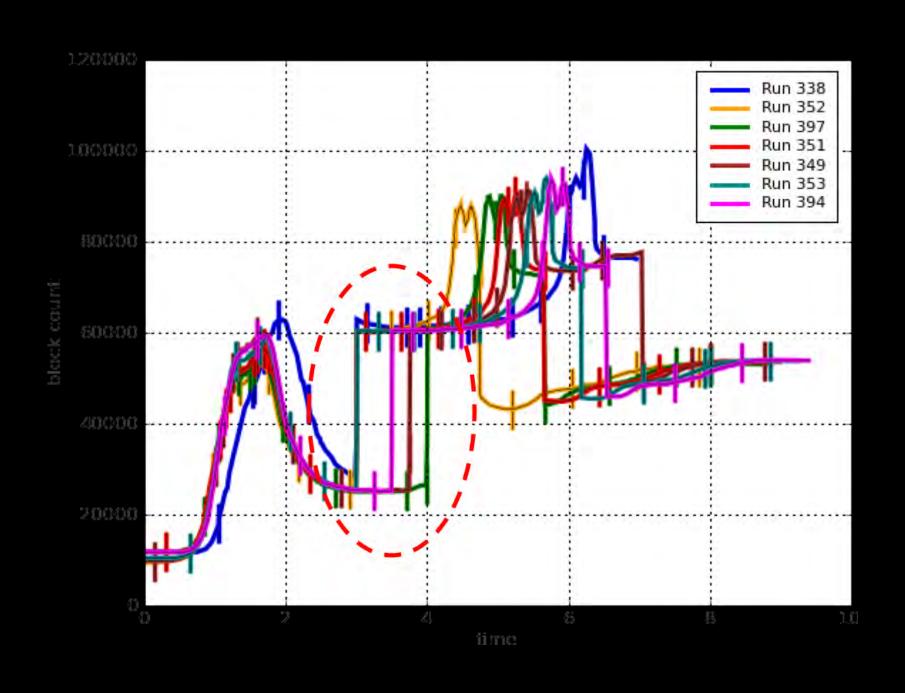
Summary View



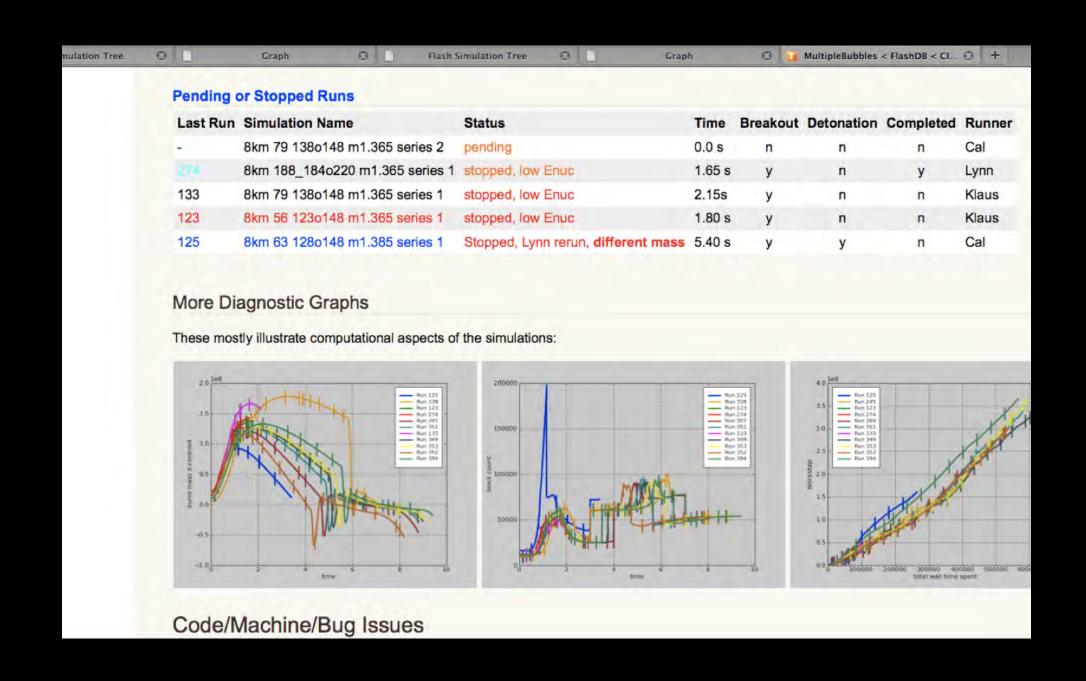
Smaash Outcomes (Simulation State)



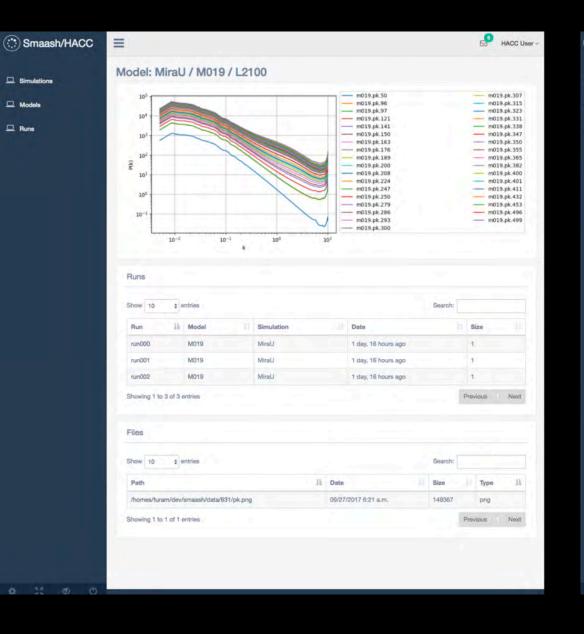
Smaash Outcomes (Analysis)

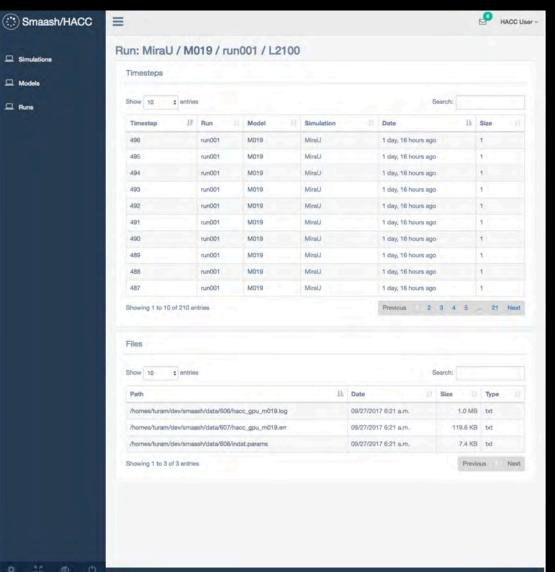


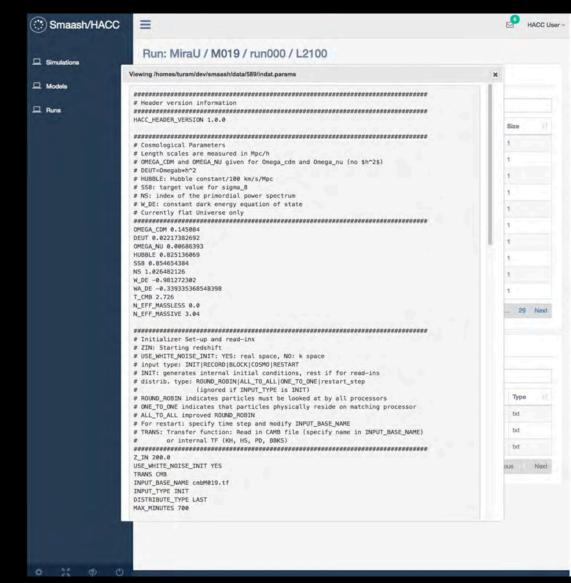
Smaash Outcome (Notebook)



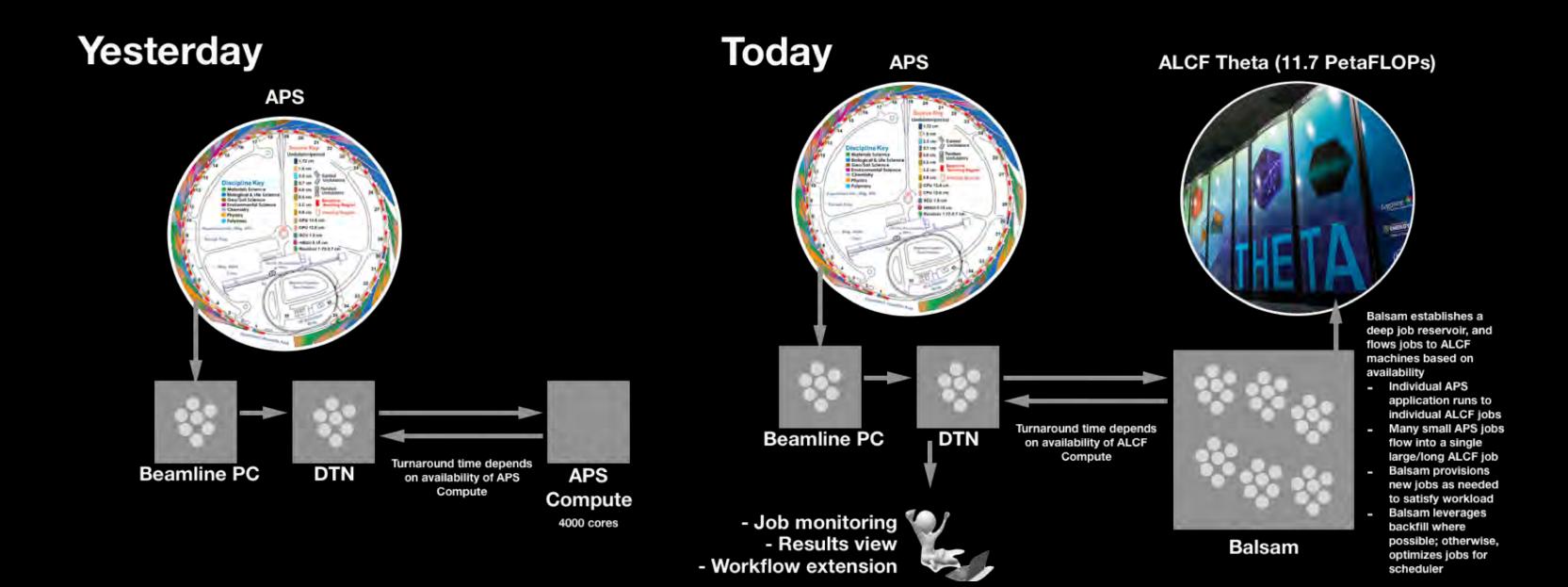
Smaash Today



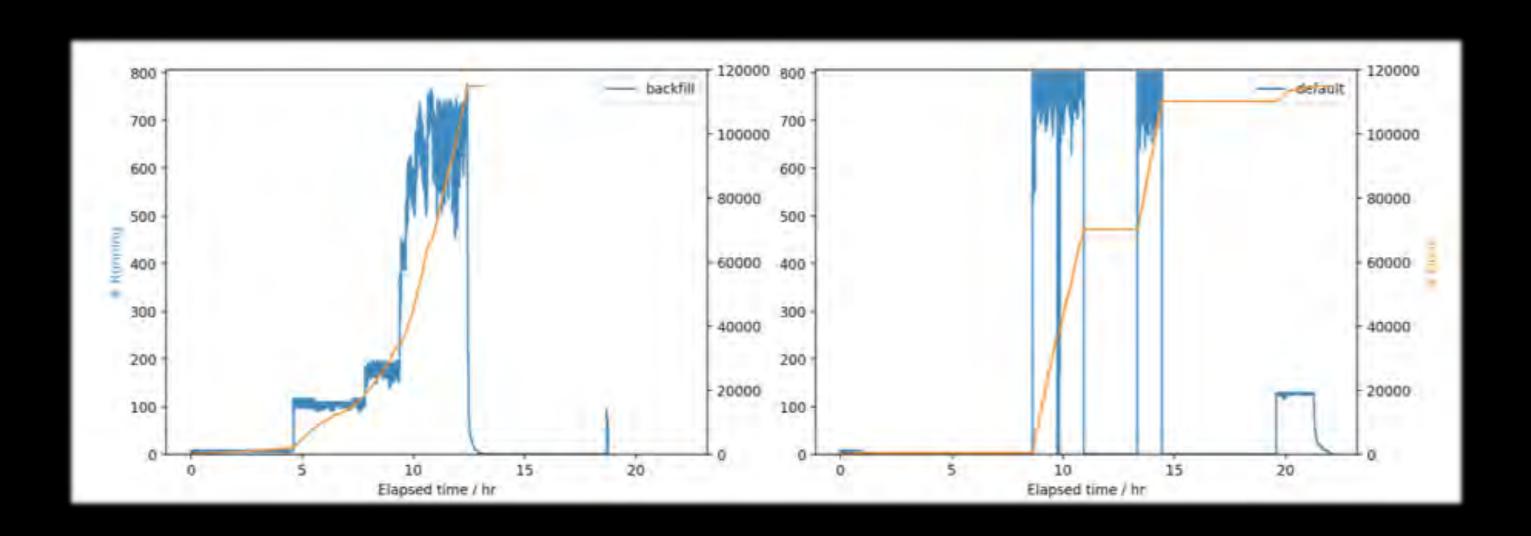




New Efforts in Science



New Efforts in Science^u



^uM. A. Salim, T. D. Uram, J. T. Childers, P. Balaprakash, V. Vishwanath and M. E. Papka, *Balsam: Automated Scheduling and Execution of Dynamic, Data-Intensive* HPC Workflows, to appear 1st Annual Workshop on Large-scale Experiment-in-the-Loop Computing, Denver, CO, November 2019.

Last Topic

Information Visualization

Connection to X science



Thank You Most of my current funding is provided by the Argonne Leadership Computing Facility a DOE Office of Science User Facility supported under contract DE-AC02-06CH11357 with additional support from the National Science Foundation. Thanks to all the students of the ddiLab and my colleagues at NIU and

ANL.

If I have seen further it is by standing on the shoulders of giants.

— Sir Isaac Newton

Extra Sides

HPC Landscape Yesterday

Simulation Applications
64bit floating point
memory bandwith
random access to memory
sparse matrices
distributed memory jobs
synchronous input/output multinode
scalability limited communication
low latency high bandwidth
large coherency domains (sometimes)
output typically greater than input
output rarely read
output is data

HPC Landscape (Today)

Simulation Applications	Big Data Applications	Deep Learning Applications
64bit floating point	64bit and integer important	lower precision <= 32bit
memory bandwith	data analysis pipelines	inferencing can be 8bit (TPU)
random access to memory	databases including NoSQL	scaled integer possible
sparse matrices	MapReduce/SPARK	training dominates development
distributed memory jobs	millions of jobs	inference dominates pro
synchronous input/output multinode	input/output bandwidth limited	reuse of training data
scalability limited communication	data management limited	data pipelines needed
low latency high bandwidth	many task parallelism	dense float point typical SGEMM small DFT, CNN
large coherency domains (sometimes)	large-data in and large-data out	ensembles and search
output typically greater than input	input and output both important	single models small
output rarely read	output is read and used	input more important than output
output is data	output is data	output is models