



Towards Real-Time Detection and Tracking of Spatio-Temporal Features: Blob-Filaments in Fusion Plasma

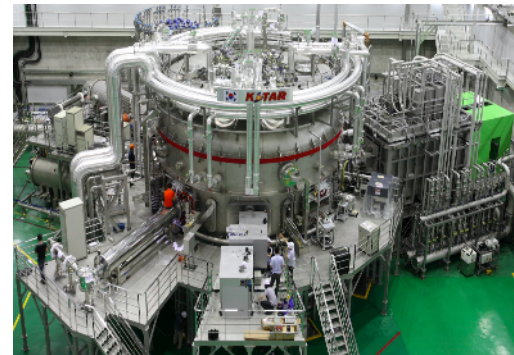


February 27, 2018

John Wu

LBNL

<http://crd.lbl.gov/sdm/>



Outline

□ Summary

- ✧ ICEE project
- ✧ Application examples

□ Data and Process Management

- ✧ ADIOS, Streaming, Subsetting, dynamic execution

□ New Feature Extraction algorithm

- ✧ Blob detection algorithm, two-level parallelization

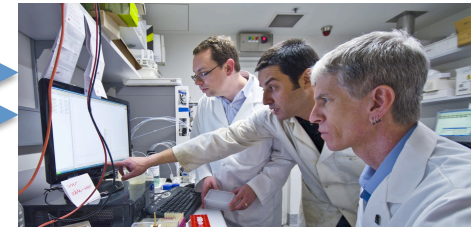
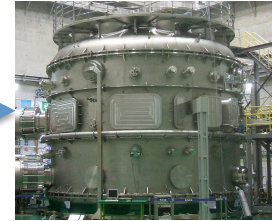
□ Fusion plasma stability

- ✧ Comparing experiment with simulation



ICEE Project Vision: Enable Real-Time Collaborative Decision Making

- ❑ Vision: Enable distributed, collaborative, real-time decisions
 - ✧ Workflows including both experiments and simulations
 - ✧ Reduce cost, improve utilization of expensive experimental devices
- ❑ Metrics of Success:
 - ✧ Reduction of time to make a “good” decision, across the entire scientific process
 - ✧ Adoption of technology by “important users”



Motivating Example: Fusion

- ❑ Complex DOE experiments, such as a fusion reactor, contain numerous diagnostics that need **Near-Real-Time analysis for feedback to the experiment**
 - ✧ For guiding the experiment
 - ✧ For faster and better understanding of the data
- ❑ Current techniques to write, read, transfer, and analyze “files” require a long time to produce an answer
 - ✧ Long delay due to slow disks involved to store files
 - ✧ Slow start up of many workflow execution engines

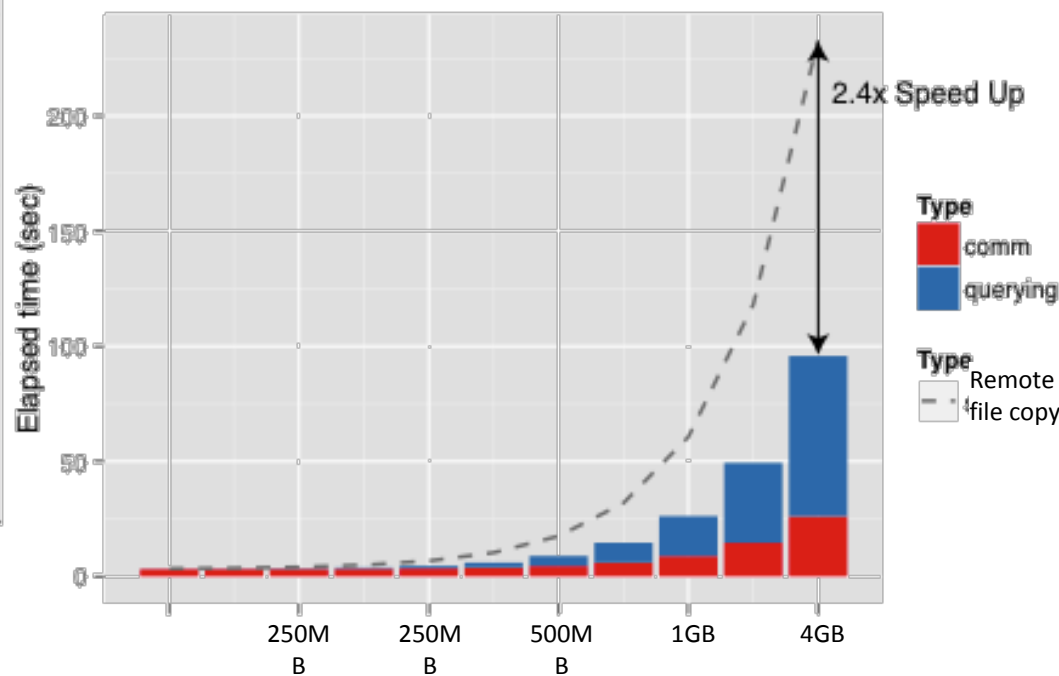
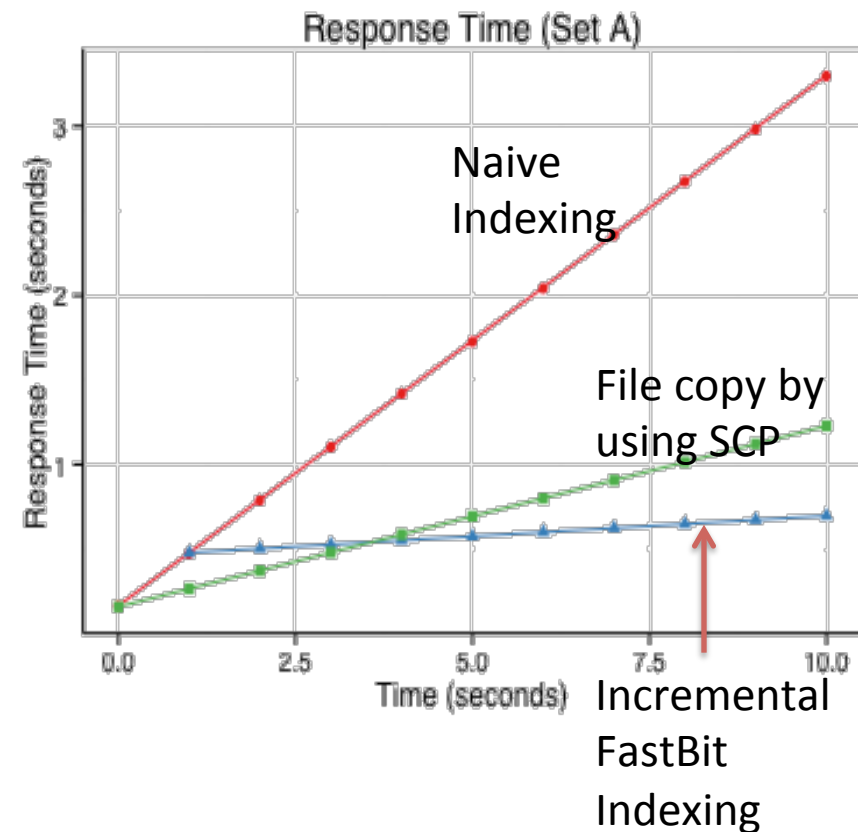
ICEE Approach

- ❑ Create an I/O abstraction layer for
 - ✧ Writing data quickly on exa, peta, tera, giga scale resources **transparently**
 - ✧ **Streaming** data on these resources, and across the world
- ❑ Place different parts of a workflow at **different locations**
- ❑ Research new techniques for quickly **indexing** data to reduce the amount of information moved in the experimental workflow
 - ✧ Prioritize data
- ❑ Create new techniques to identify important **features**, which turn the workflow into a data-driven streaming workflow

Index-and-Query Reduces Execution Time

❑ Remote file copy VS. index-and-query

- ✧ Measured between LBNL and ORNL to simulate KSTAR-LBNL-ORNL connection
- ✧ Indexed by FastBit. Observed a linear performance (i.e., indexing cost increased by data size) → Expensive indexing cost
- ✧ However, once we have index built, index-and-query can be a better choice over remote file copy



Use Case I: Near Real Time Detection of Blobs

❖ Fusion Plasma blobs

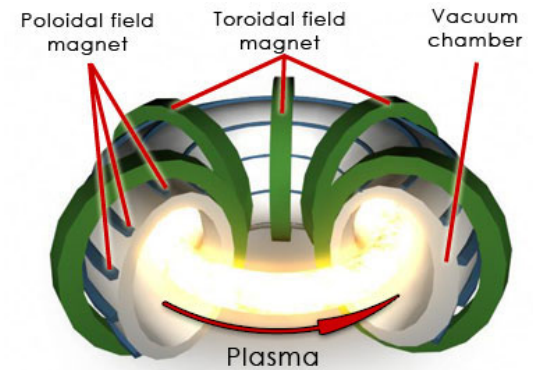
- ❖ Lead to the loss of energy from tokamak plasmas
- ❖ Could damage multi-billion tokamak

❖ The experimental facility may not have enough computing power for the necessary data processing

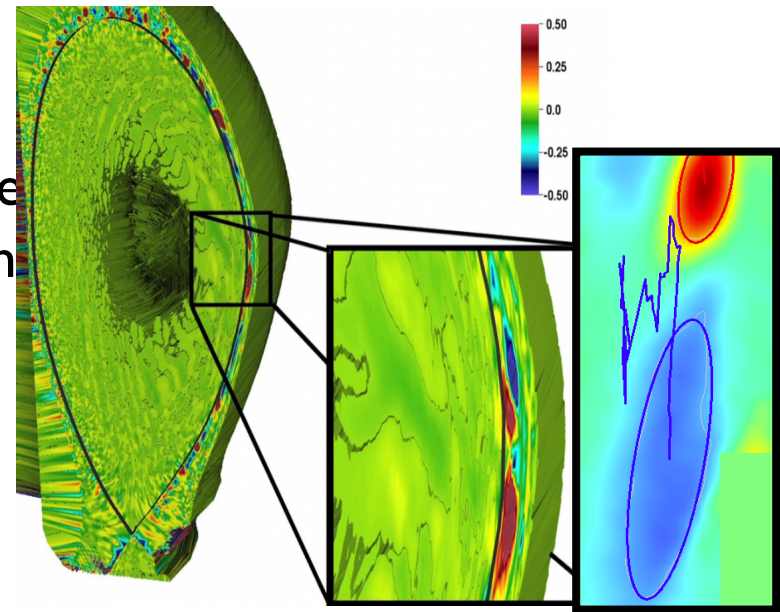
❖ Distributed in transient processing

- ❖ Make more processing power available
- ❖ Allow more scientists to participate in the data analysis operations and monitor the experiment remotely
- ❖ Enable scientists to share knowledge and processes

❖ Lingfei Wu, Alex Sim, Jong Choi, M. Churchill, K Wu, S Klasky, CS Chang



© 2005 HowStuffWorks

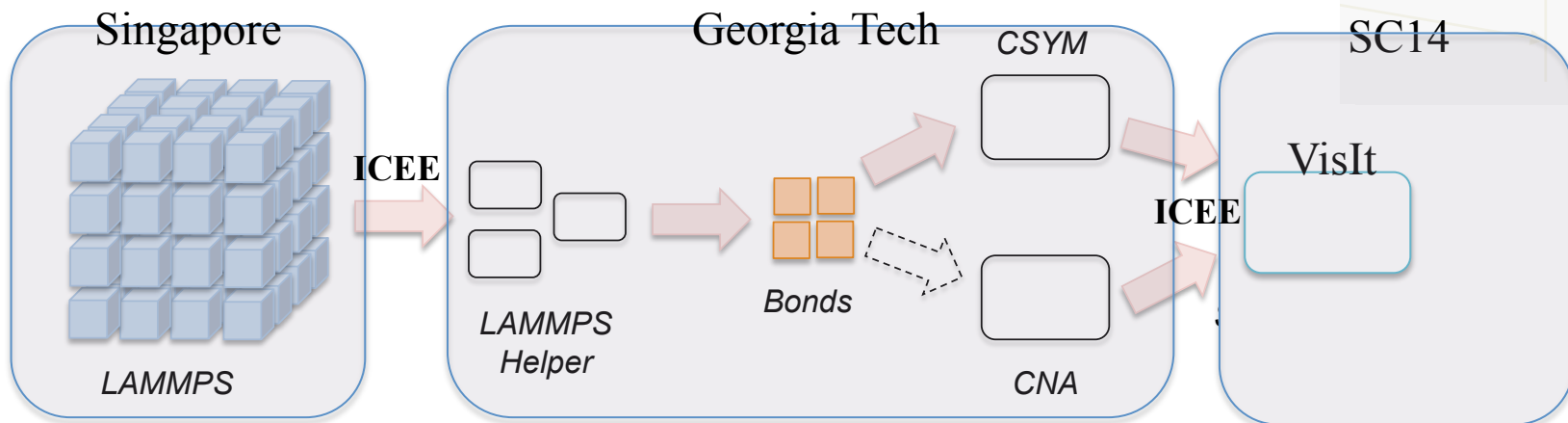
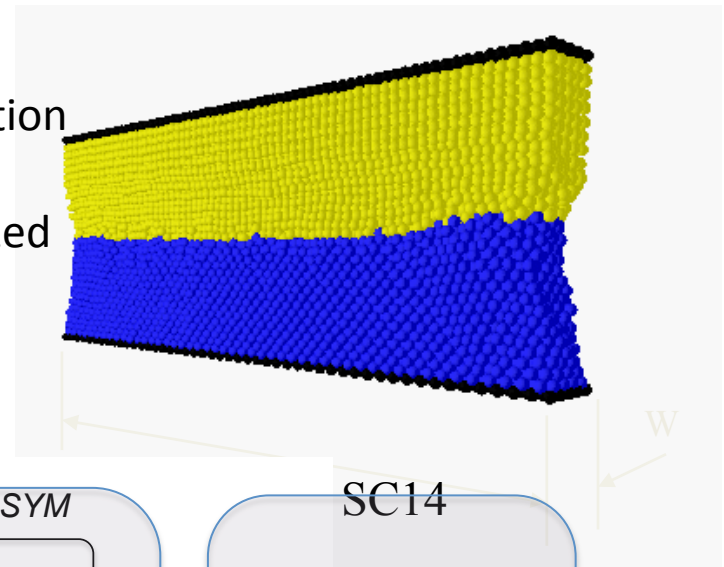


Blobs in fusion reaction Blob trajectory
(Source: EPSI project)

Use Case 2: Fracture of Nano-Materials

Matthew Wolf, Jai Dayal, and Greg Eisenhauer, Georgia Tech

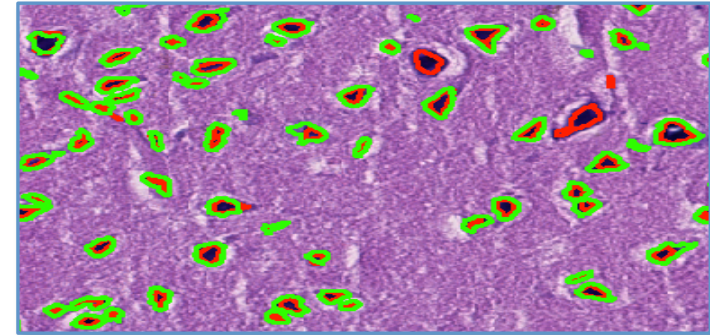
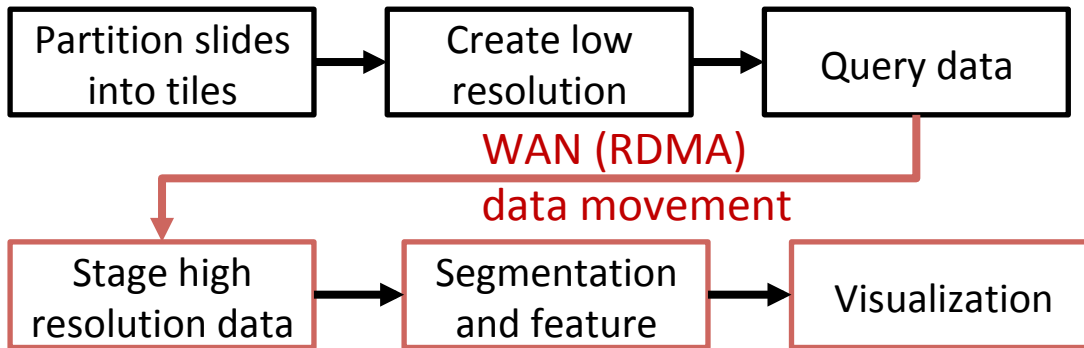
- This demonstration is based off a scenario from materials scientists interested in understanding fracture in nano-structured materials
- It uses LAMMPS to simulate the block of nano-structured metal while under stress.
- Simulation proceeds until the first plastic deformation (start of fracture) is detected.
- At that first fracture, the system is fully characterized to understand where and, hopefully, why things broke.



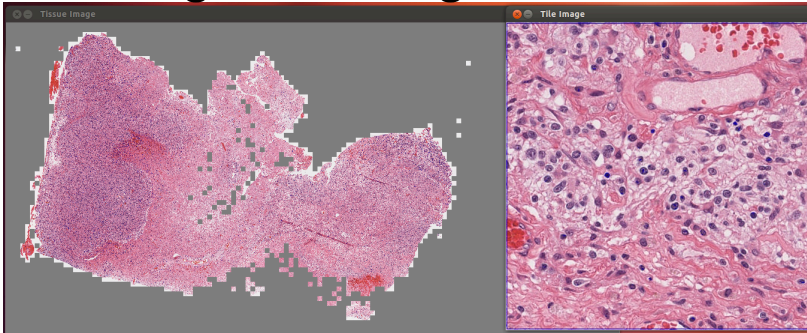
Use Case 3: Microscopy Image Analysis

J. Saltz, T. Kurc, M. Michalewicz, M. Parashar + ICEE team

- ❑ **Significance:** Understanding of disease morphology at micro - anatomic level has potential for better diagnosis disease mechanisms.
- **Challenge:** Rapidly analyze tissue slides (120Kx120K pixels) to assess condition



- **Technologies:** (1) SBU ADIOS for wide-area, efficient transfers; (2) Longbow for very fast, low-latency connection; (3) pipelined processing on clusters
- **Demo:** Tissue slides on machine in Singapore. Analysis done on cluster at Georgia Tech. Segmentation results displayed on client machine.



Snapshot of adaptive processing of a remote slide (53Kx36K pixel resolution).

JS, TK, MM & MP supported by NCI and NLM: 1U24CA180924, R01LM011119-01, R01LM009239 grants, and ORNL SDAV and JFA.

Overview: Enable Rapid Decision Making

❑ Effective data management

- ✧ Easily express data accesses: high-level data model instead of offsets into files
- ✧ Transparent accesses to remote data
- ✧ Convenient subsetting operations

❑ Effective workflow management

- ✧ Tight integration of workflow components to reduce latency
- ✧ Make the best uses of known resources

❑ Reduce the time to solution

- ✧ Streaming data accesses, avoid waiting for all data before analysis could start
- ✧ Only access the necessary data records (selective data accesses)
- ✧ Keep the data in memory as much as possible (*in situ* processing)

Main Tasks of ICEE

- ❑ Create an infrastructure that transparently
 - ✧ Stage data used in workflows on local nodes
 - ✧ Stage data used in workflows on remote nodes
 - ✧ Stage data through files, using an external file mover
 - ✧ Index the data and move only the relevant chunks of data from the query
- ❑ Dynamically adjust the data being moved according to
 - ✧ Rules the user provides
 - ✧ Dynamic changes in the networking and computational resources
 - ✧ Multiple workflows being run concurrently
- ❑ Efficient merging of multiple data streams
 - ✧ Enable comparative analytics

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□ New Feature Extraction algorithm

- ✧ Blob detection algorithm, two-level parallelization

□ Fusion plasma stability

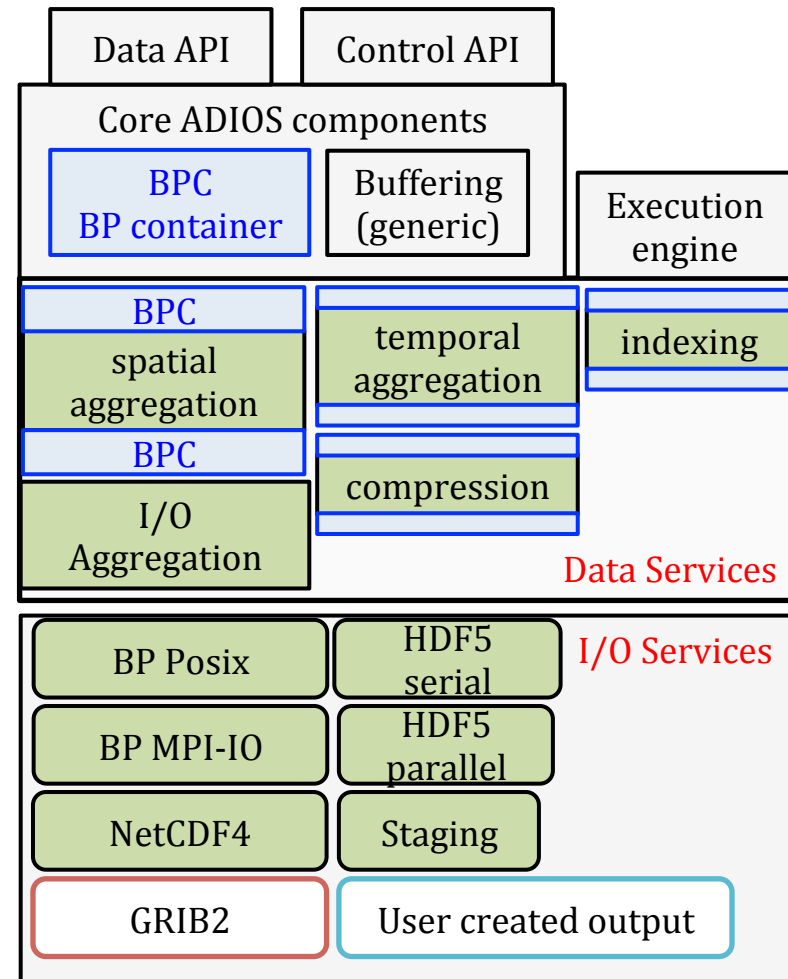
- ✧ Comparing experiment with simulation



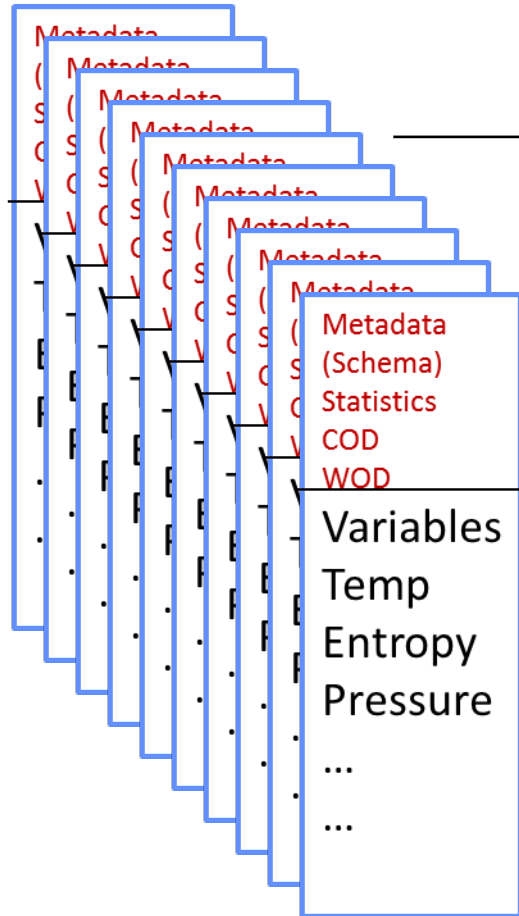
ADIOS Abstraction Unifies Local And Remote I/O

- ❑ I/O Componentization for Data-at-Rest and Data-in-Motion
- ❑ Service Oriented Architecture for Extreme scaling computing
- ❑ Self Describing data movement/storage
- ❑ Main paper to cite

Q. Liu, J. Logan, Y. Tian, H. Abbasi, N. Podhorszki, J. Choi, S. Klasky, R. Tchoua, J. Lofstead, R. Oldfield, M. Parashar, N. Samatova, K. Schwan, A. Shoshani, M. Wolf, K. Wu, W. Yu, "Hello ADIOS: the challenges and lessons of developing leadership class I/O frameworks", Concurrency and Computation: Practice and Experience, 2013



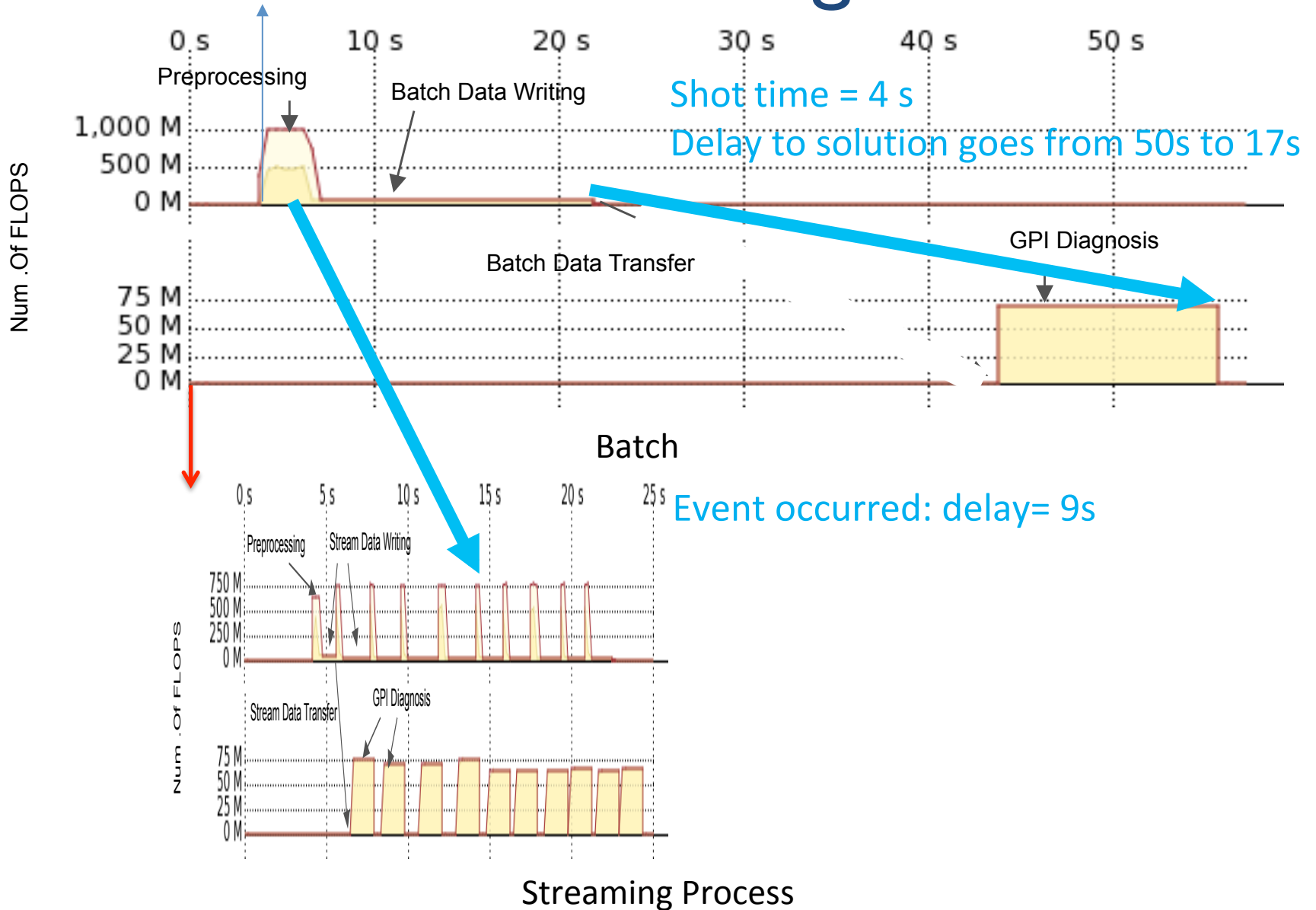
The ADIOS-BP Stream/File format



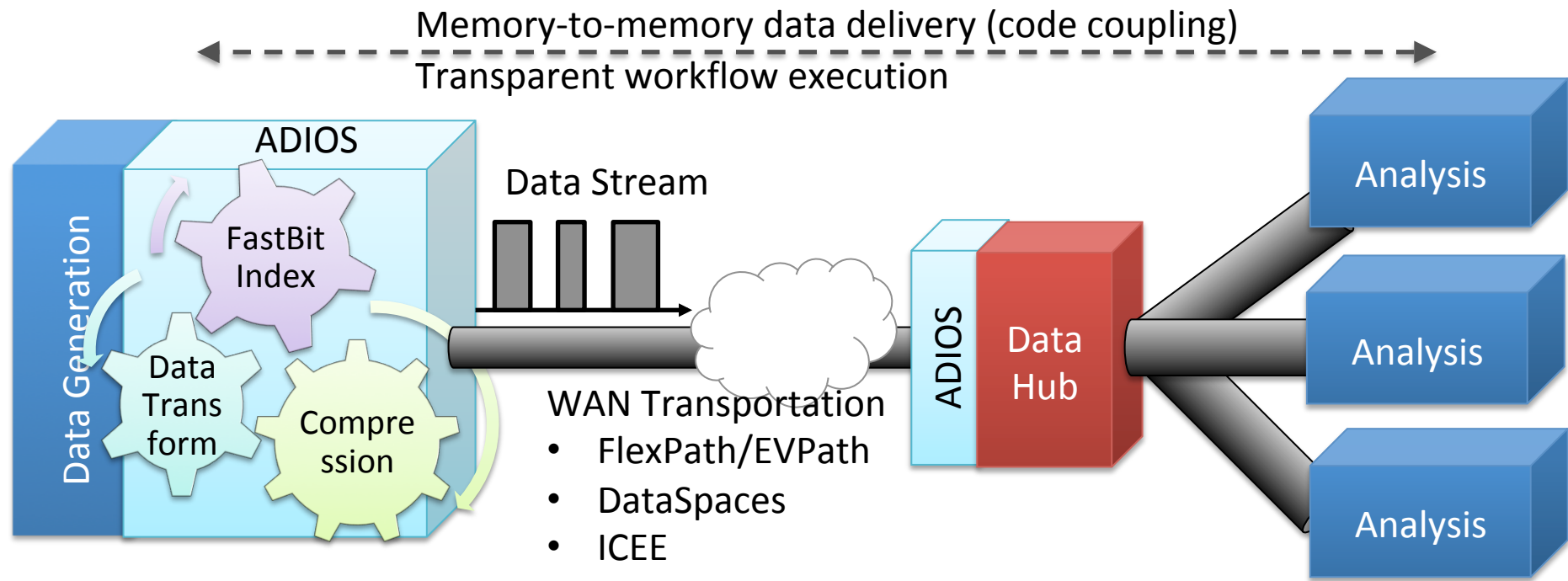
- ❑ All data chunks are from a single producer
 - ✧ MPI process, Single diagnostic
- ❑ Ability to create a separate metadata file when “sub-files” are generated
- ❑ Allows variables to be individually compressed
- ❑ Has a schema to introspect the information
- ❑ Has workflows embedded into the data streams
- ❑ Format is for “data-in-motion” and “data-at-rest”

Ensemble of chunks = file

Batch Vs. Streaming Process



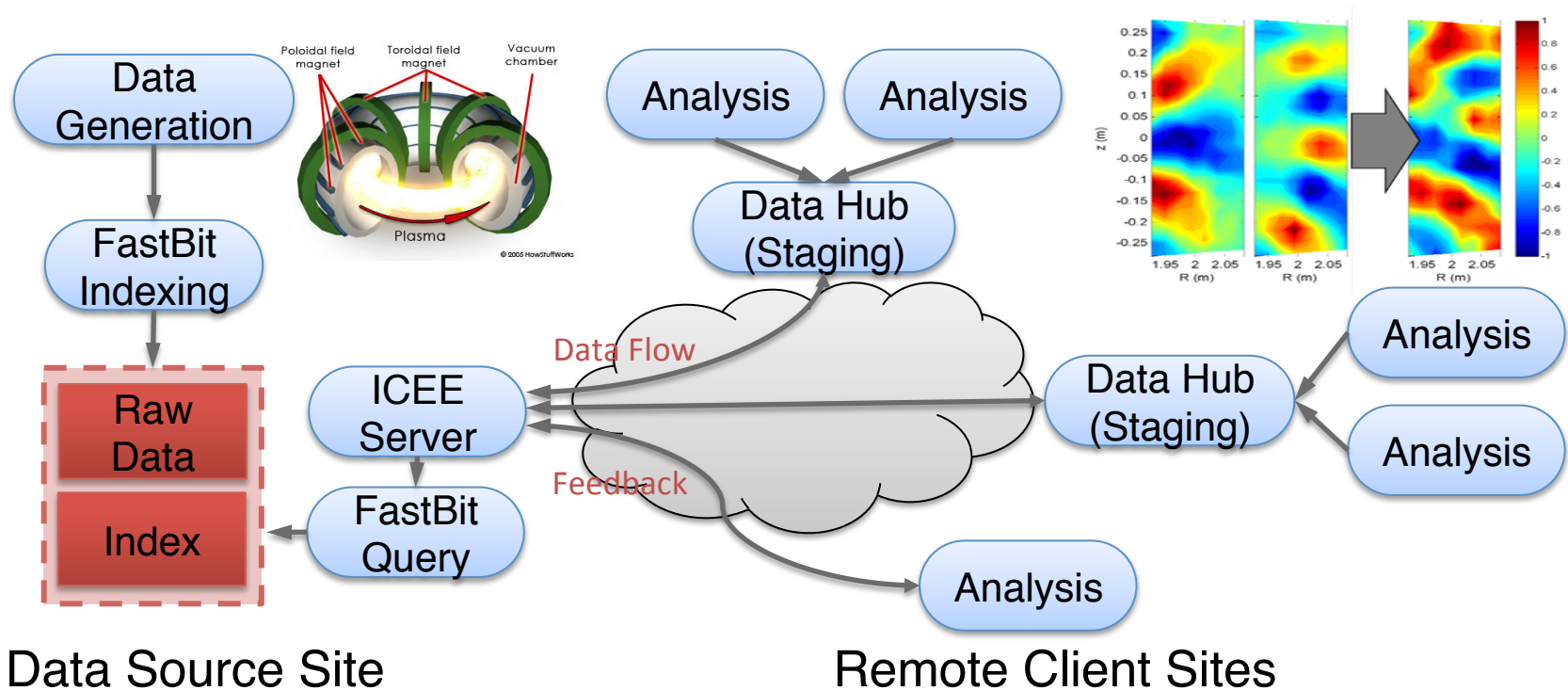
ICEE Enables Distributed In Memory Workflows



Research on stream-based WAN data process

- ✧ In-transit processing (supporting data-in-memory)
- ✧ Data indexing & query to reduce network payload
- ✧ WAN transportation: FlexPath (GATech), DataSpaces (Rutgers), ICEE (ORNL/LBNL)

ICEE System Development



□ Features

- ADIOS provides an overlay network to share data and give feedbacks
- Stream data processing – supports stream-based IO to process pulse data
- In transit processing – provides remote memory-to-memory mapping between data source (data generator) and client (data consumer)
- Indexing and querying with FastBit technology

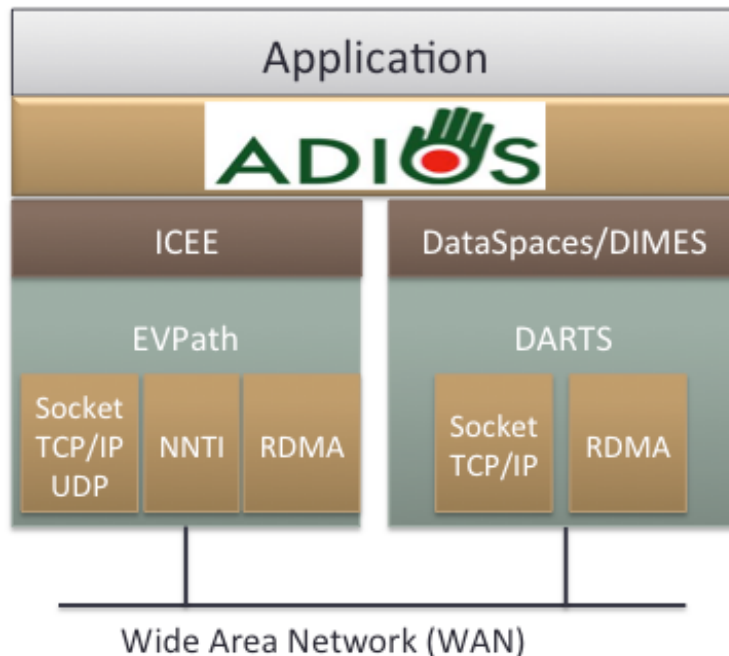
Software Components of ICEE Transport

❑ ICEE

- ✧ Using EVPath package (GATech)
- ✧ Support uniform network interface for TCP/IP and RDMA
- ✧ Easy to build an overlay network

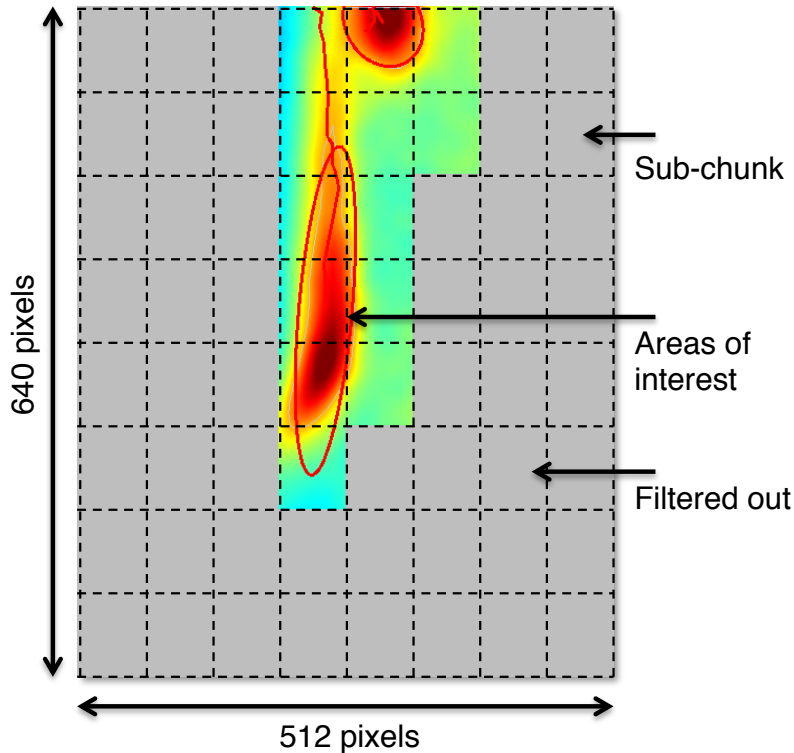
❑ Dataspaces (with sockets)

- ✧ Developed by Rutgers
- ✧ Support TCP/IP and RDMA



The screenshot shows the ADIOS website header with the Oak Ridge National Laboratory logo and navigation links: HOME, ABOUT OLCF, LEADERSHIP SCIENCE, COMPUTING RESOURCES, CENTER PROJECTS, and SUPPORT. The main content area is titled 'Adios' and includes a navigation bar with links for Overview, Download & Revision History, Documents & Manuals, and Press & Publications. Below this, there is a section for 'Download' with the current stable version ADIOS 1.9.0 and a link to download it. A 'Revision History' section follows, detailing the 1.9.0 release from December 2015 and the 1.8.0 release from December 2014. The 1.8.0 release section is highlighted with a blue box and lists features such as the Query API, Staging over WAN, and various bug fixes.

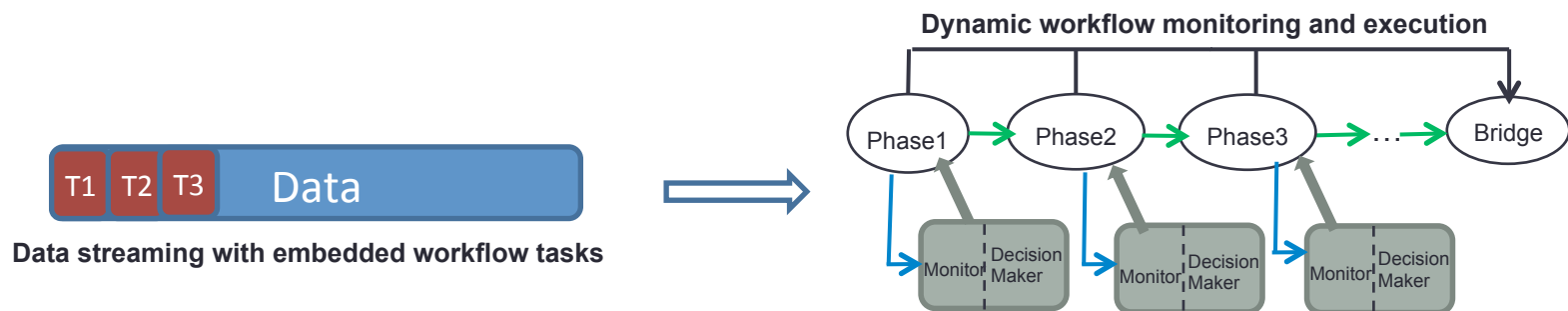
Reducing Payload Size



- ❑ Select only areas of interest and send (e.g., blobs)
- ❑ Reduce payload on average by about 5X

Adaptive Workflow Execution

- ❑ Adaptively determine *what* phases to perform, *where* by embedding workflow tasks with data streams
- ❑ The execution runtime needs provide management for
 - Dynamic allocation of resources for global optimization without strong consistency in the knowledge across the entire system
 - Adaptation policy and mechanisms within the runtime for changing power and performance metrics
 - Isolation of faults and minimization of interference of the entire system
 - Flexible reconfiguration of the workflow to support rapid evolution of user and application requirements



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□ **New Feature Extraction algorithm**

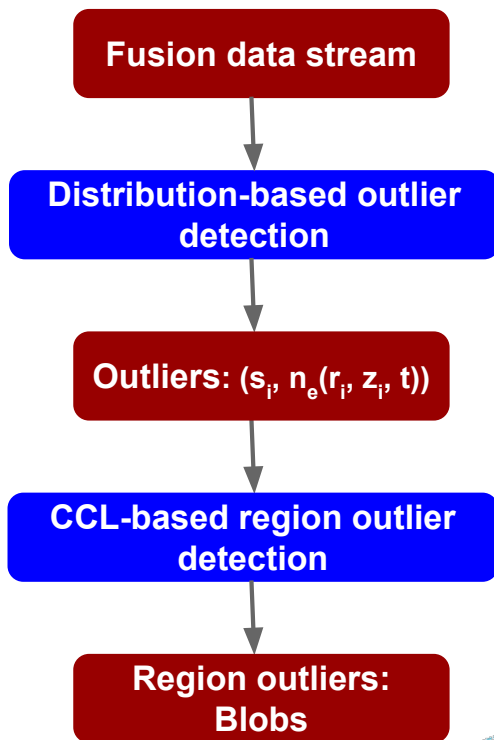
- ✧ **Blob detection algorithm, two-level parallelization**

□ Fusion plasma stability

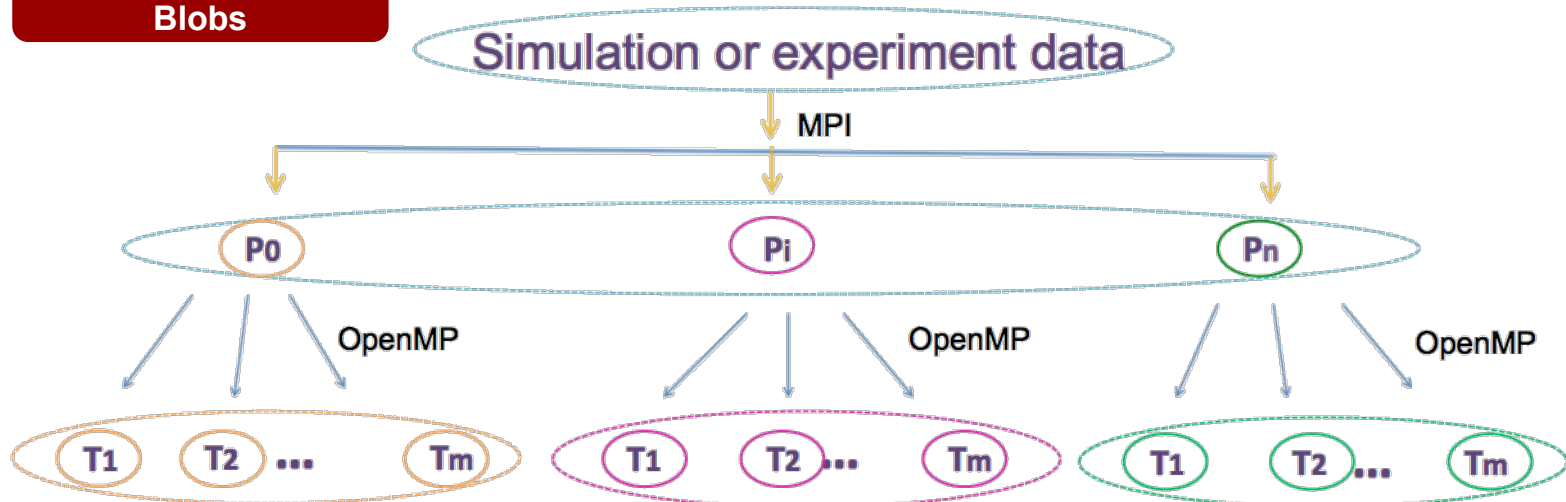
- ✧ Comparing experiment with simulation



Outline of Feature Detection Algorithm



- ❑ Formulate the blob detection problem as a region outlier detection problem
- ❑ Develop a high-performance approach to meet the real-time requirements
- ❑ A hybrid MPI/OpenMP parallelization on many-core processor architecture:
- ❑ High-level: use MPI to allocate n processes and each process takes at least one time frame
- ❑ Low-level: use OpenMP to accelerate the computations with m threads



Spatial Feature Extraction Approach

- ❑ Target: regions of interest defined on range conditions on known quantities, e.g., “temperature between 800 and 1000 and pressure less 10^5 ”
- ❑ Use database indexing technology to resolve the conditions, which generally identifies “points” satisfying the conditions
- ❑ Connect the points into regions in space

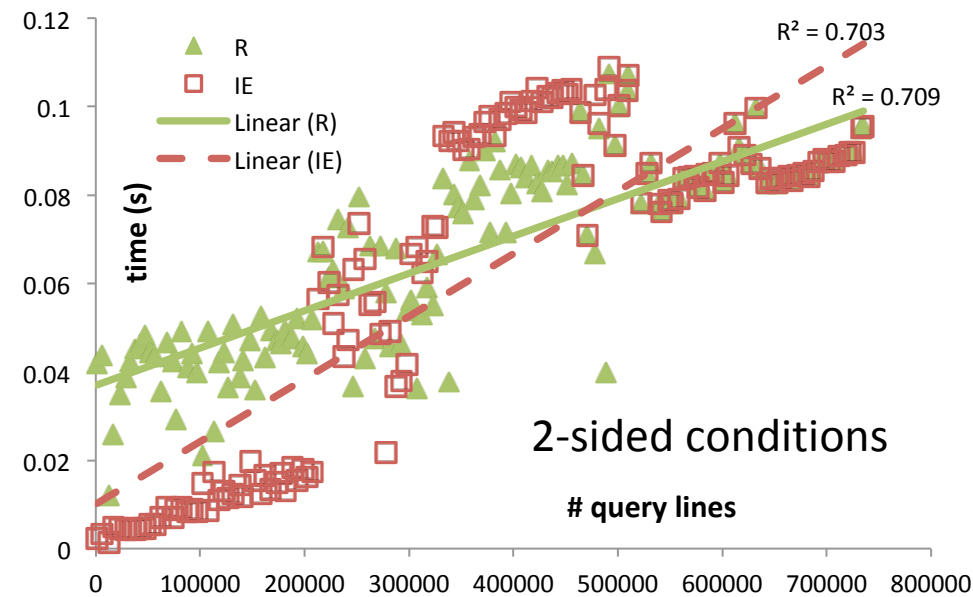
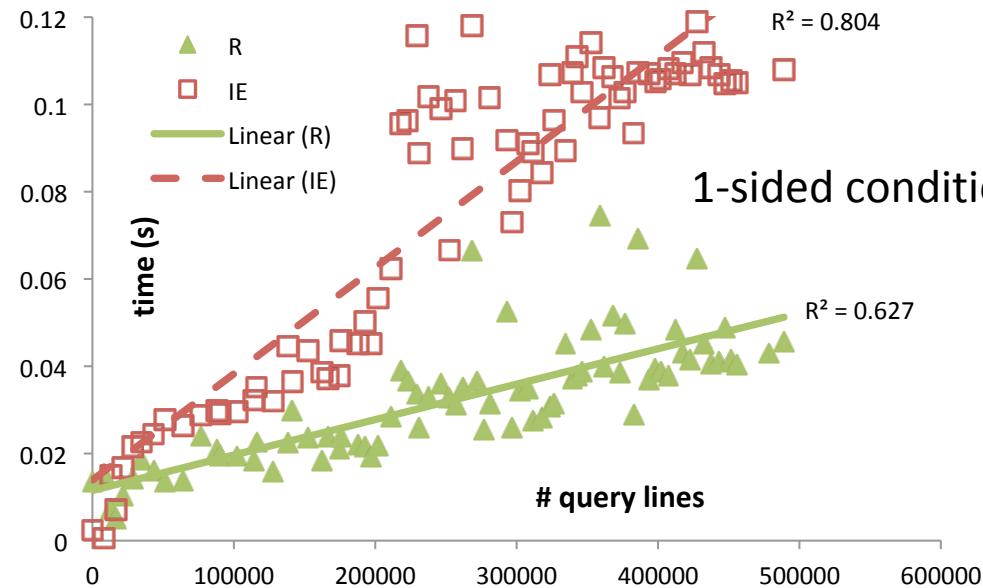
Operations on Indexes

- ❑ Review available database indexing technologies
 - ✧ Known multi-dimensional indexes suffer from “curse of dimensionality” – don’t work for high dimensional data
 - ✧ Most common indexes, e.g., B-tree, don’t work well for ad hoc conditions – require too many combinations of variables to be indexed
 - ✧ Compressed bitmap index supports ad hoc queries and works well for high dimensional data – our favorite
- ❑ Surprise: Let N denote the number of points in the dataset, V denote the volume of the regions of interest, and S denote the surface of the regions
 - ✧ Previous results show that find V points takes $O(V)$ time
 - ✧ Our tests show $O(S)$ time! (Note that $S < V$, often much less)

Connected Component Labeling

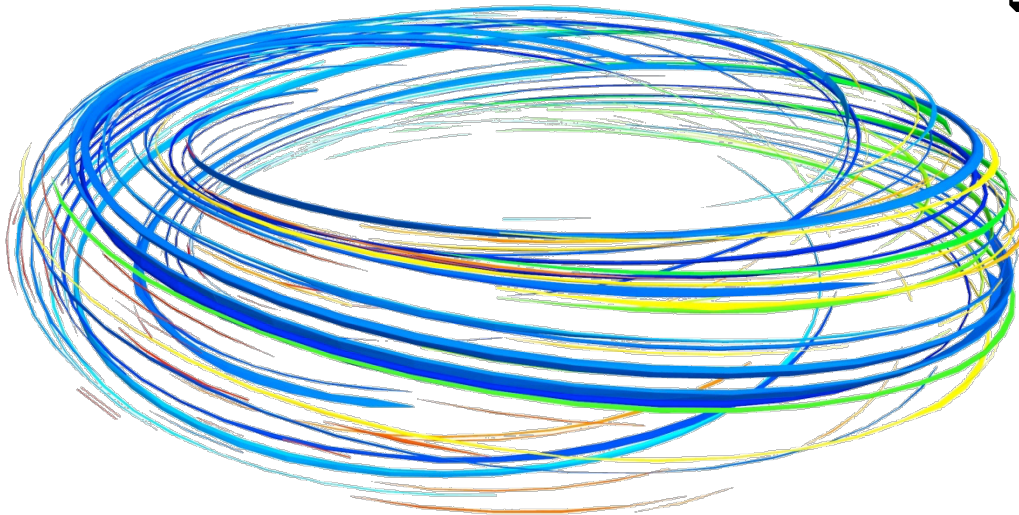
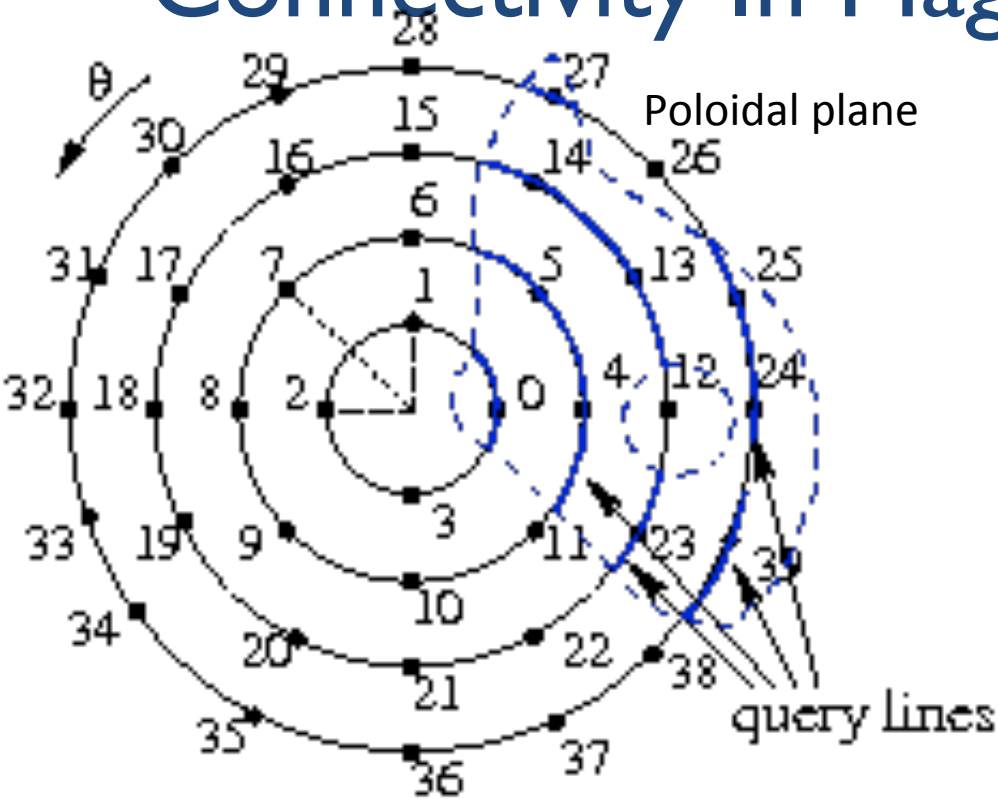
- ❑ Connect points into regions with connected component labeling algorithms
- ❑ Our contributions:
 - ✧ Represent the connectivity in an very efficient manner using magnetic coordinates
 - Makes it much easier to find which neighbors are connected to each other, **reduce execution time by hundreds of times**
 - ✧ Use an efficient connected component labeling algorithm named Scan with Array-based Union-Find (SAUF)
 - SAUF requires less memory than alternatives and is generally faster as well
 - ✧ Use a compact representation of the points in the regions of interest named query lines
 - Reduces the execution time significantly because the number of query lines are much less than the number of points

Bitmap Index Operation Time



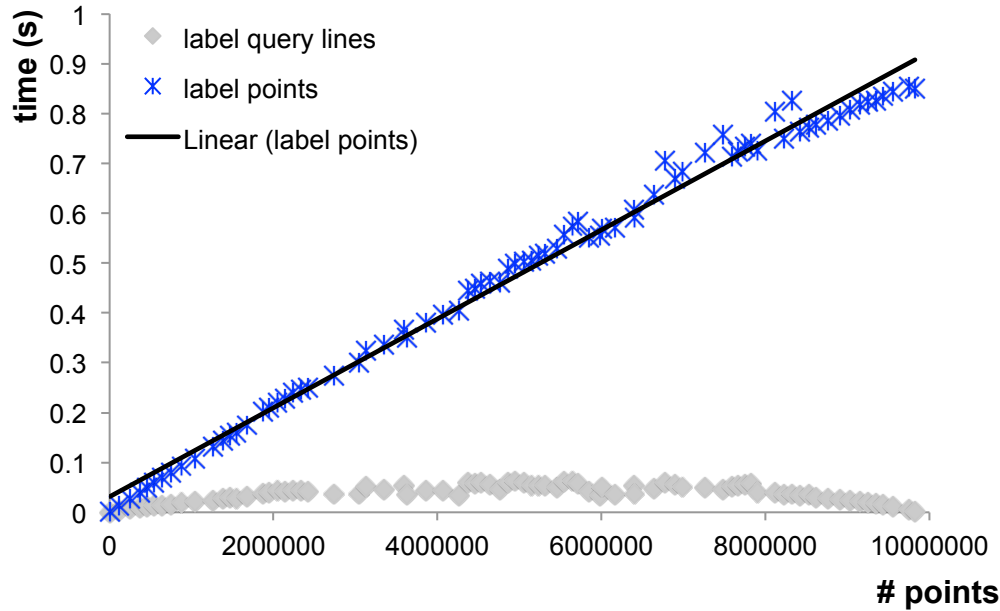
- Let Q denote the number of query lines
- For 1-dimensional 1-sided range conditions, e.g., “pressure $\geq 10^5$,” the range indexes (R) takes $O(Q)$ time, but can be very large in size
- The new Interval-Equality index can be much smaller, but take 3X as long to resolve the same conditions
- $Q \leq S$

Connectivity In Magnetic Coordinates



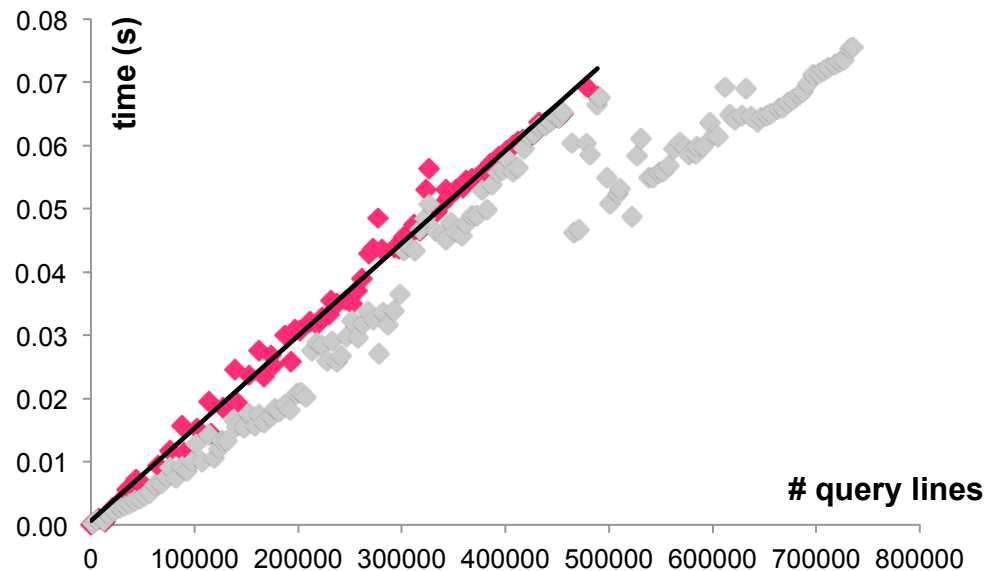
- ❑ Much more compact than the general connectivity graph: ~ 200 numbers vs. 6 million numbers
- ❑ Follows the physics used for the simulation (Gyrokinetic Toroidal Code)
- ❑ Much cleaner connectivity definition: a point only connects to
 - ✧ Two points on the same circle
 - ✧ Four points on the neighboring circles
 - ✧ Two points in the neighboring planes

Labeling Time



❑ Top figure: time to label the points is a linear function of the number of points, but $\sim 11\times$ longer than labeling the query lines

❑ Bottom figure: time to label the query lines is bounded by a linear function of the number of query lines (i.e., $O(Q)$): red points from 1-sided range conditions and gray from 2-sided range conditions



❑ Labeling query lines using magnetic coordinates is $600-1000\times$ faster than using connectivity graph

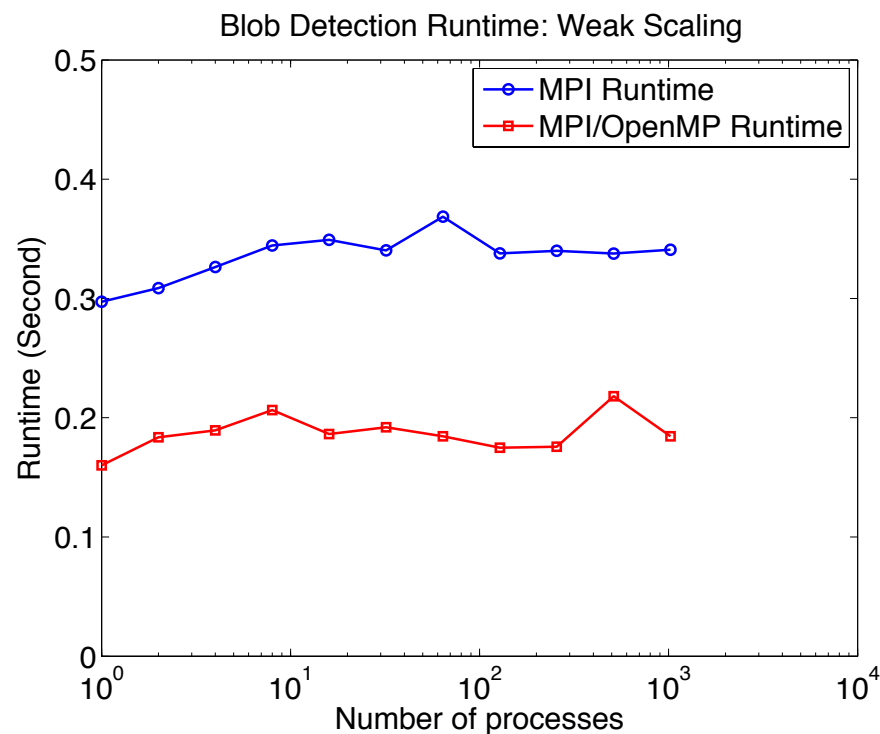
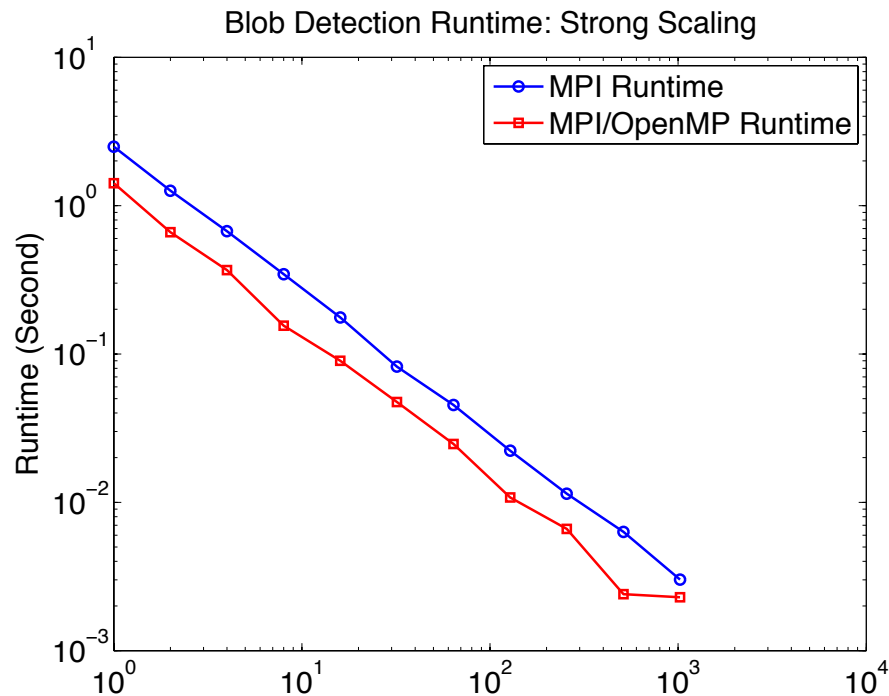
Real-time Blob Detection

❖ Top right figure: strong scaling

- ❖ Complete blob detection in around 2 ms with MPI/OpenMP using 4096 cores and in 3 ms with MPI using 1024 cores
- ❖ Linear time speedup in blob detection time
- ❖ MPI/OpenMP is two times faster than MPI

❖ Bottom right figure: weak scaling

- ❖ Near constant blob detection time indicates our implementations scale very well to solve much larger problems



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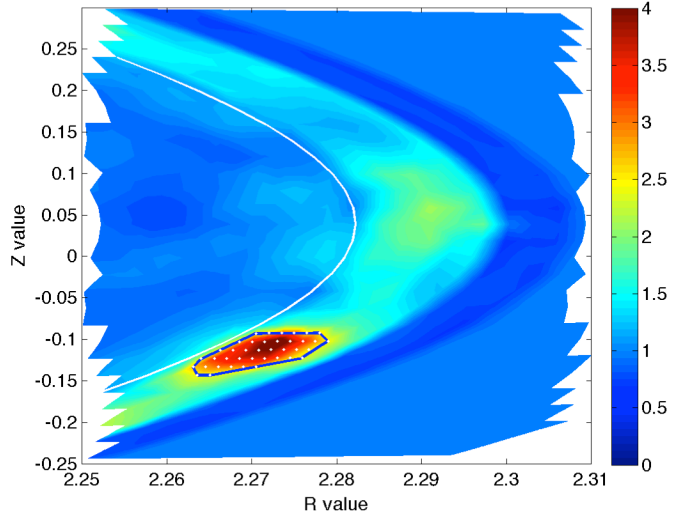
Fusion Example with More Details

- ❑ **Volume:** Initially 90 TB per day, 18 PB per year, maturing to 2.2 PB per day, 440 PB per year
- ❑ **Value:** All data are taken from expensive instruments for valuable reasons.
- ❑ **Velocity:** Peak 50 GB/s, with **near real-time analysis needs**
- ❑ **Variety:** ~100 different types of instruments and sensors, numbering in the thousands, producing interdependent data in various formats
- ❑ **Veracity:** The quality of the data can vary greatly depending upon the instruments and sensors.

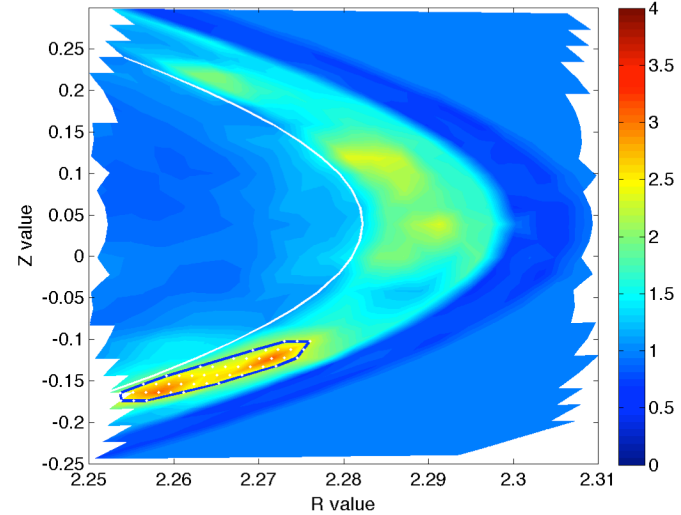
The pre-ITER superconducting fusion experiments outside of US will also produce increasingly bigger data (KSTAR, EAST, Wendelstein 7-X, and JT60-SU later).

Cross Sections of Hot Plasma in Torus

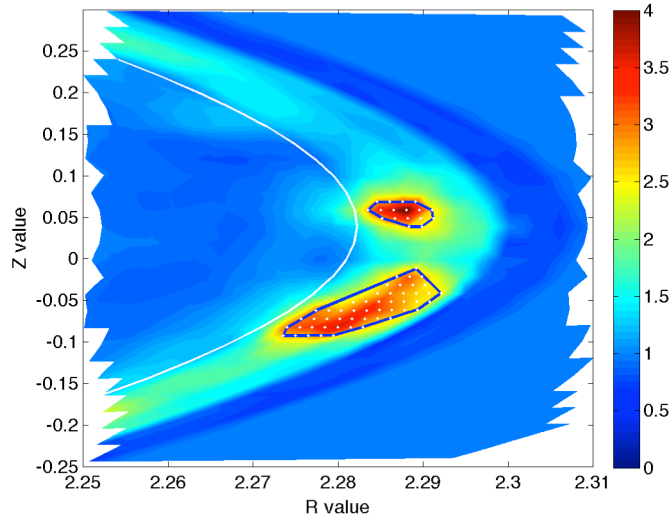
Blob Detection: time frame 86 and Poloidal plane ζ



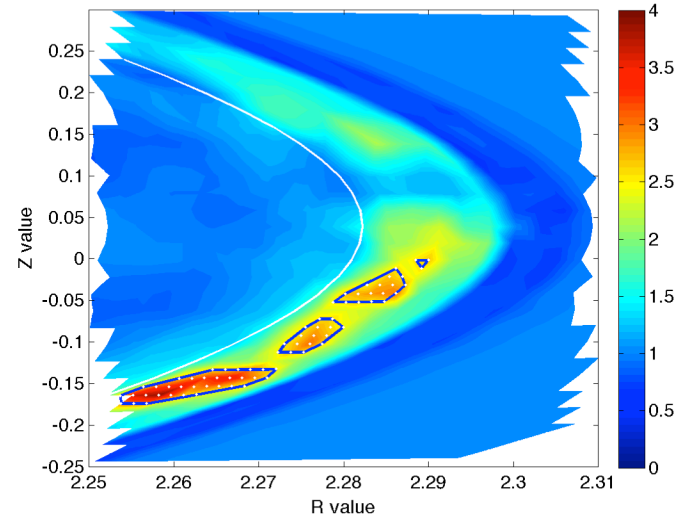
Blob Detection: time frame 86 and Poloidal plane ξ



Blob Detection: time frame 86 and Poloidal plane 1



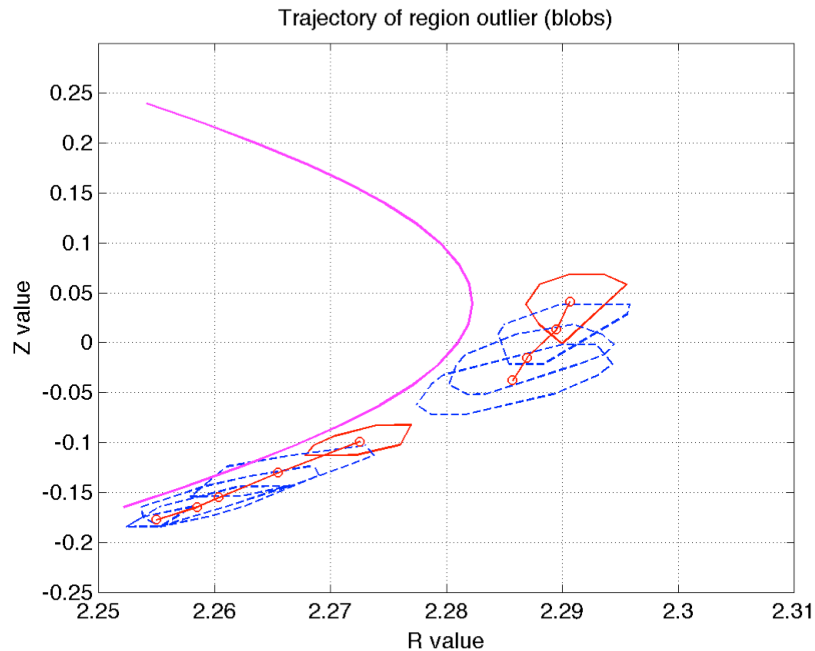
Blob Detection: time frame 86 and Poloidal plane ζ



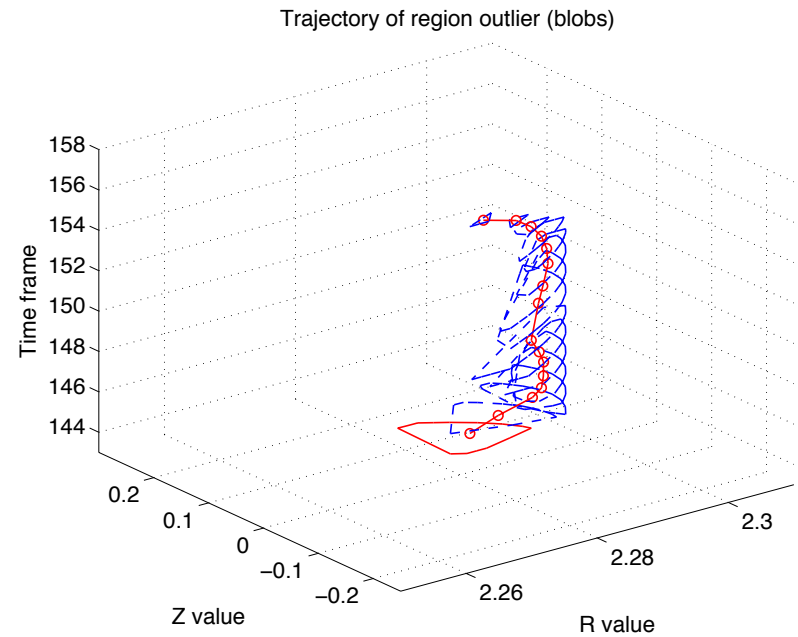
Blobs in red

Tracking Trajectories

2D examples showing multiple regions (i.e., blobs)



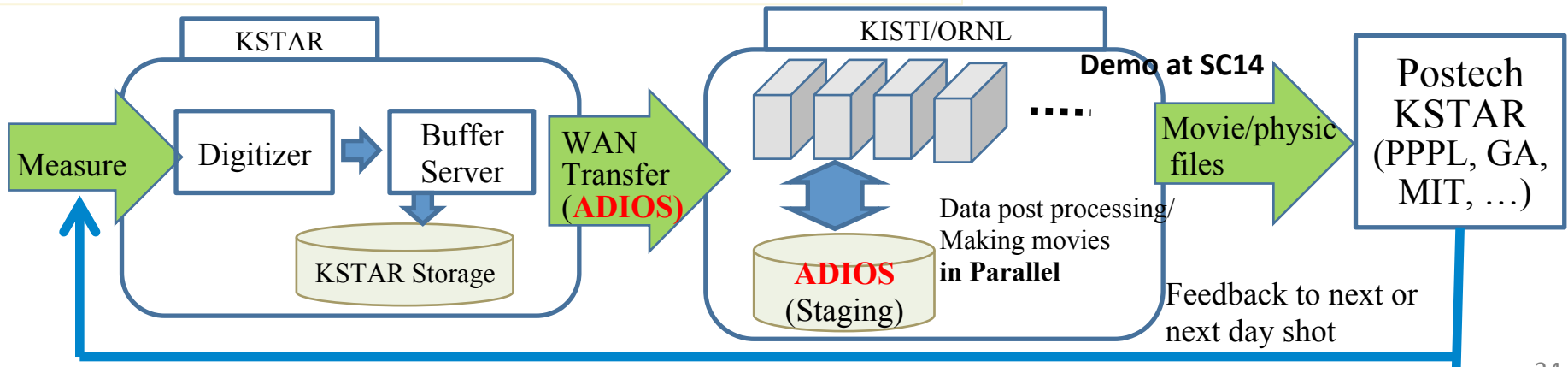
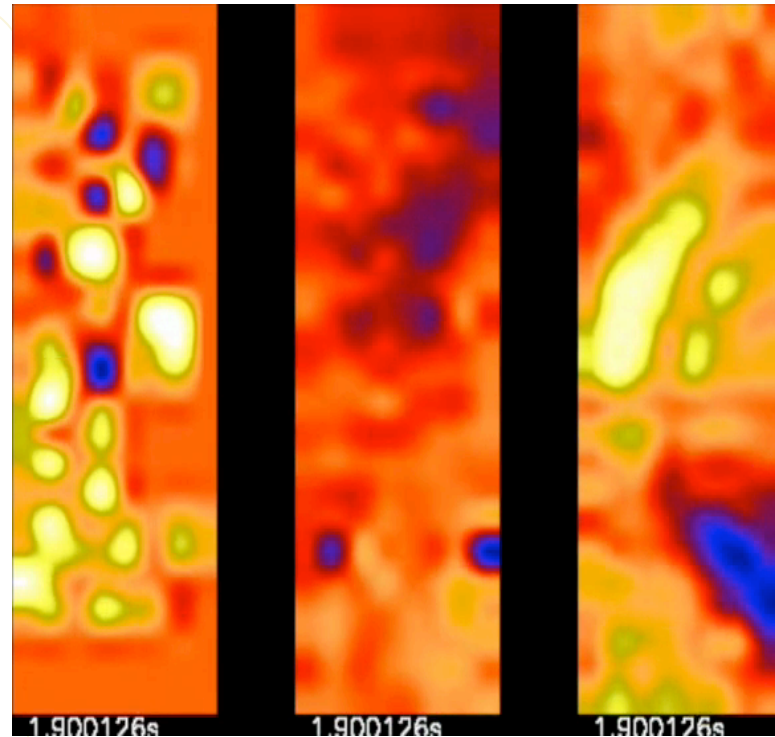
3D example showing a single region (blob) over 15 time steps



Review: Distributed Streaming

KSTAR ECEI Image Analysis Workflow

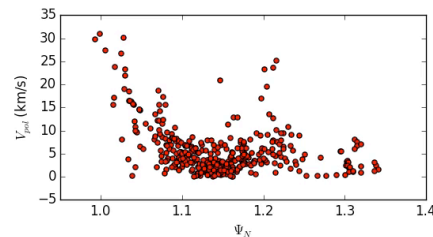
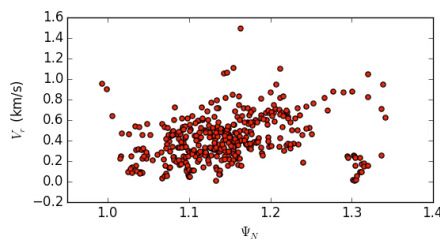
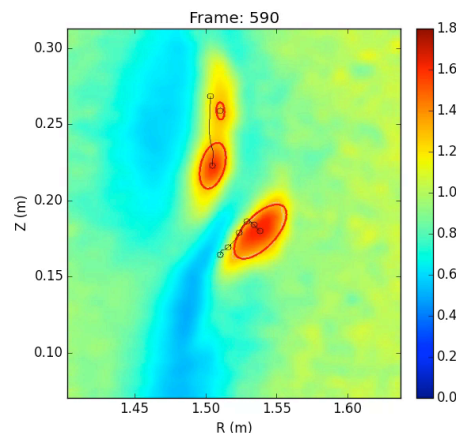
- ❑ **Objective:** To enable remote scientists to study ECE-Image movies of blobby turbulence and instabilities between experimental shots in near real-time.
- ❑ **Input:** Raw ECEi voltage data (~550MB/s, over 300 seconds in the future) + Metadata (experimental setting)
- ❑ **Requirement:** Data transfer, processing, and feedback **within <15min** (inter-shot time)
- ❑ **Implementation:** distributed data processing with ADIOS ICEE method



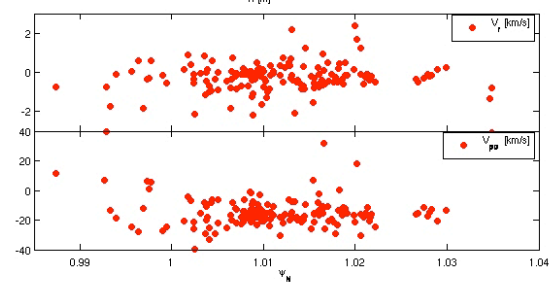
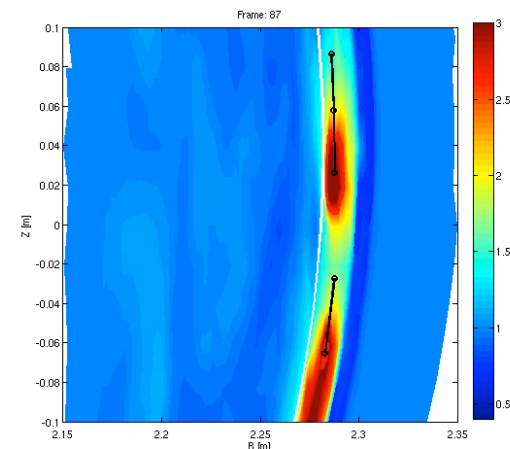
Science Use Case I: Real-time Comparison of Experiment and Simulation

- **Objective:** Enable comparisons of simulation (pre/post) and experiment at remote locations
- **Input:** Gas Puff Imaging (GPI) fast camera images from NSTX and XGCI edge simulation data
- **Output:** Blob physics
- **Requirement:** Complete in near real-time for inter-shot experimental comparison, experiment-simulation validation or simulation monitoring
- **Implementation:** distributed data processing with ADIOS ICEE method, optimized detection algorithms for near real-time analysis

Experiment
(NSTX GPI)



Simulation
(XGCI)



Lessons learned

❑ Velocity

- ✧ Critical to quickly build an index which can be done in a timely fashion

❑ Veracity

- ✧ Understand the trade-offs for accuracy (of the query) vs. accuracy of the results vs. performance (time to solution).

❑ Volume

- ✧ Reduce the volume of data being moved and processed over the WAN (size vs. accuracy)

❑ Variety

- ✧ Enable multiple streams of data to be analyzed together

❑ Value

- ✧ Provide the freedom for scientists to access and analyze their data interactively

Contact Info

- ❑ Coauthors: Lingfei Wu, Kesheng John Wu, Alex Sim, Michael Churchill, Jong Y Choi, Andreas Stathopoulos, Choong-Seock Chang, Scott Klasky
- ❑ IEEE Transactions on Big Data 2016.
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