

SCIENTIFIC DATA MANAGEMENT: Essential Technology for Data-Intensive Science

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Motivating Example 1: HENP Data Generation Analysis Phases

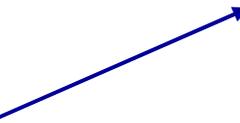
- ❖ Optimizing Storage Management and Data Access for High Energy and Nuclear Physics Applications

Experiment	# members /institutions	Date of first data	# events/year	volume/year-TB
STAR	350/35	2001	10^8 - 10^9	500
PHENIX	350/35	2001	10^9	600
BABAR	300/30	1999	10^9	80
CLAS	200/40	1997	10^{10}	300
ATLAS	1200/140	2007	10^{10}	5000

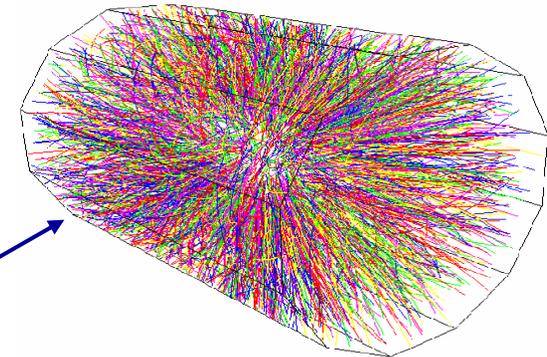


STAR: Solenoidal Tracker At **RHIC**
RHIC: Relativistic Heavy Ion Collider

LHC: Large Hadron Collider
Includes: ATLAS, STAR, ...



A mockup of
An "event"

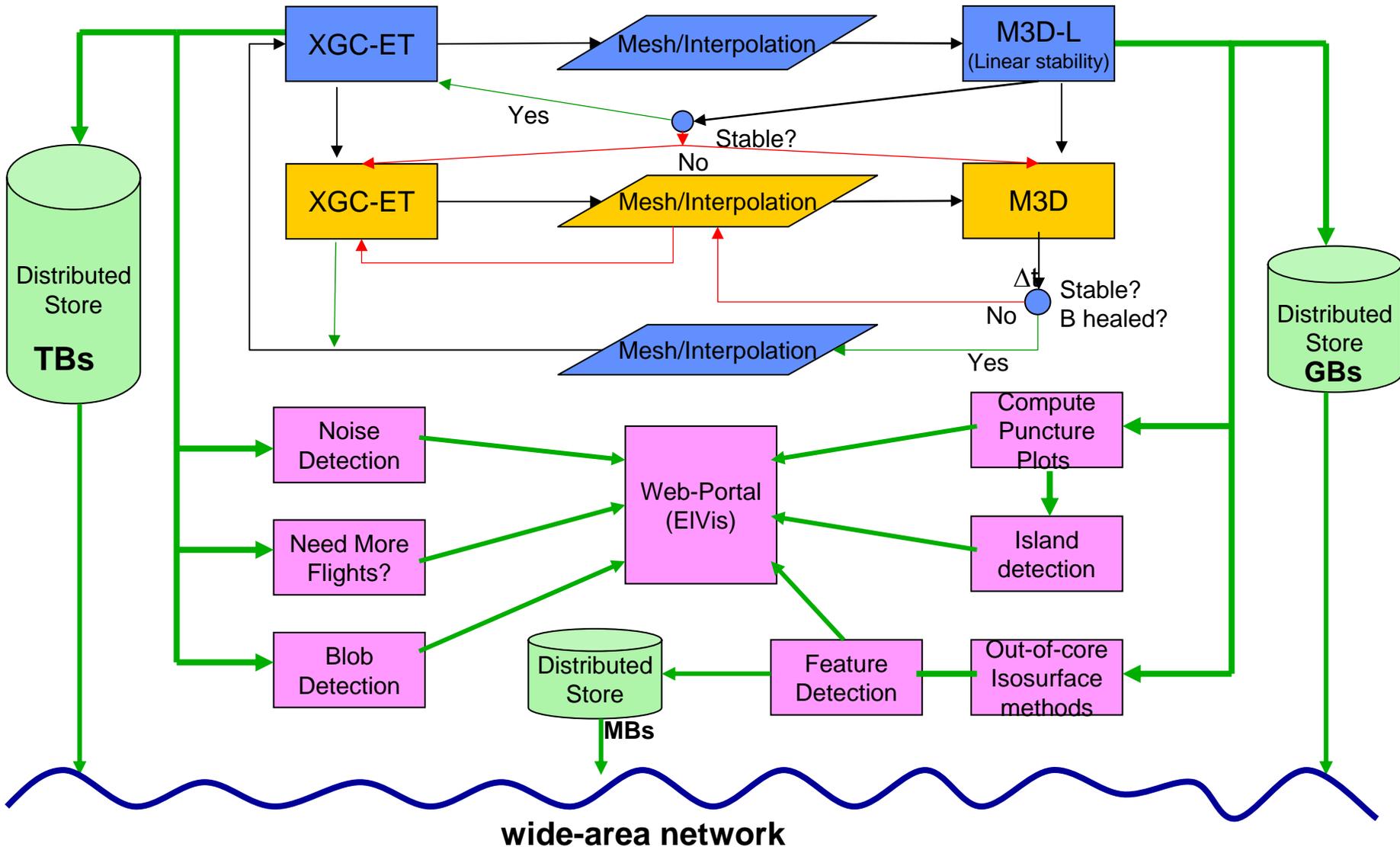


Typical Scientific Exploration Process



- ❖ Data Generation Phase
 - Generate large amounts of raw data
 - large simulations
 - collect from experiments
- ❖ Post-processing Phase
 - analyze data (find particles produced, tracks)
 - generate summary data
 - e.g. momentum, no. of pions, transverse energy
 - Number of properties is large (50-100)
- ❖ Analysis Phase
 - use summary data as guide
 - extract subsets from the large dataset
 - Need to access events based on partial properties specification (range queries)
 - e.g. $((0.1 < AVpT < 0.2) \wedge (10 < Np < 20)) \vee (N > 6000)$
 - apply analysis code

Motivating Example 3: Fusion Coordination between Running Codes



Motivating Example 2: Climate Modeling Large Scale Data Distribution

❖ Earth System Grid

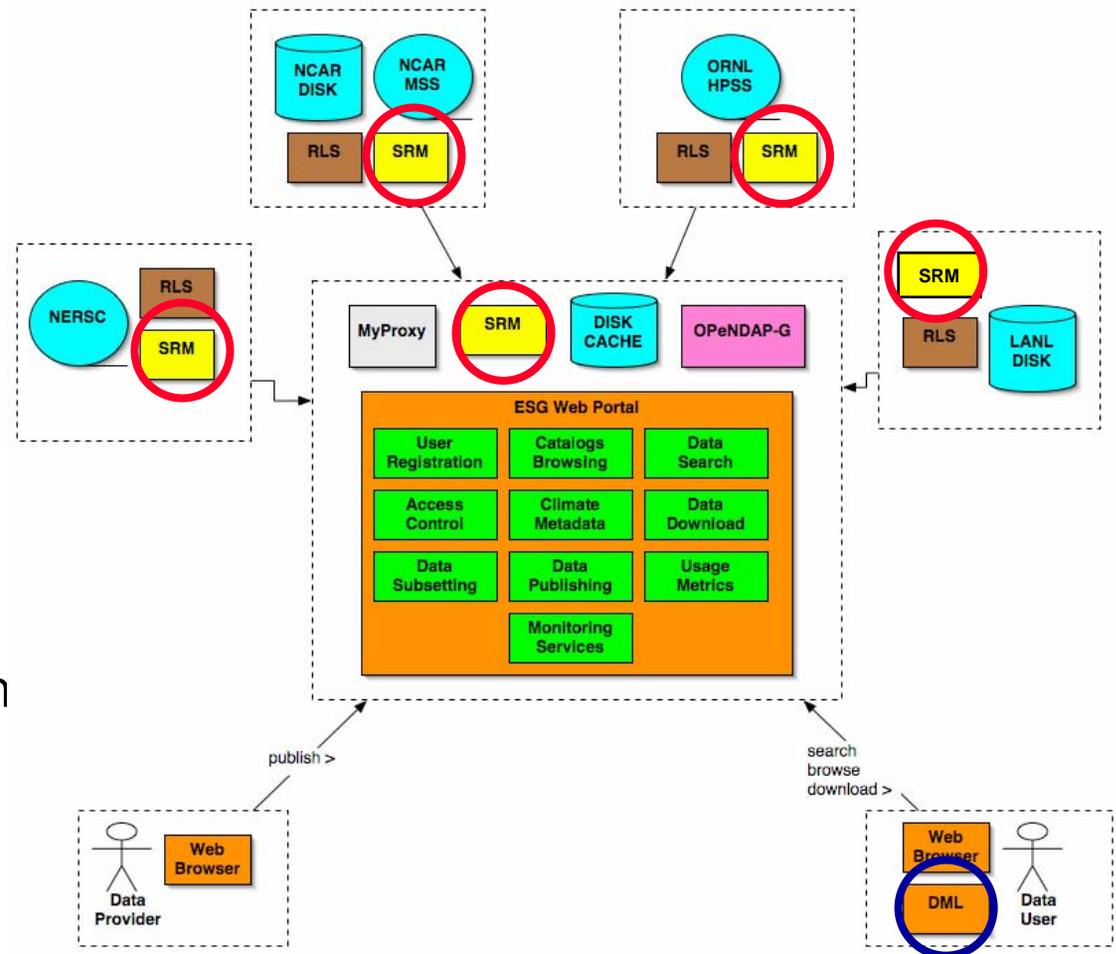
Accessing large distributed stores by 1000's of scientists

❖ Problems

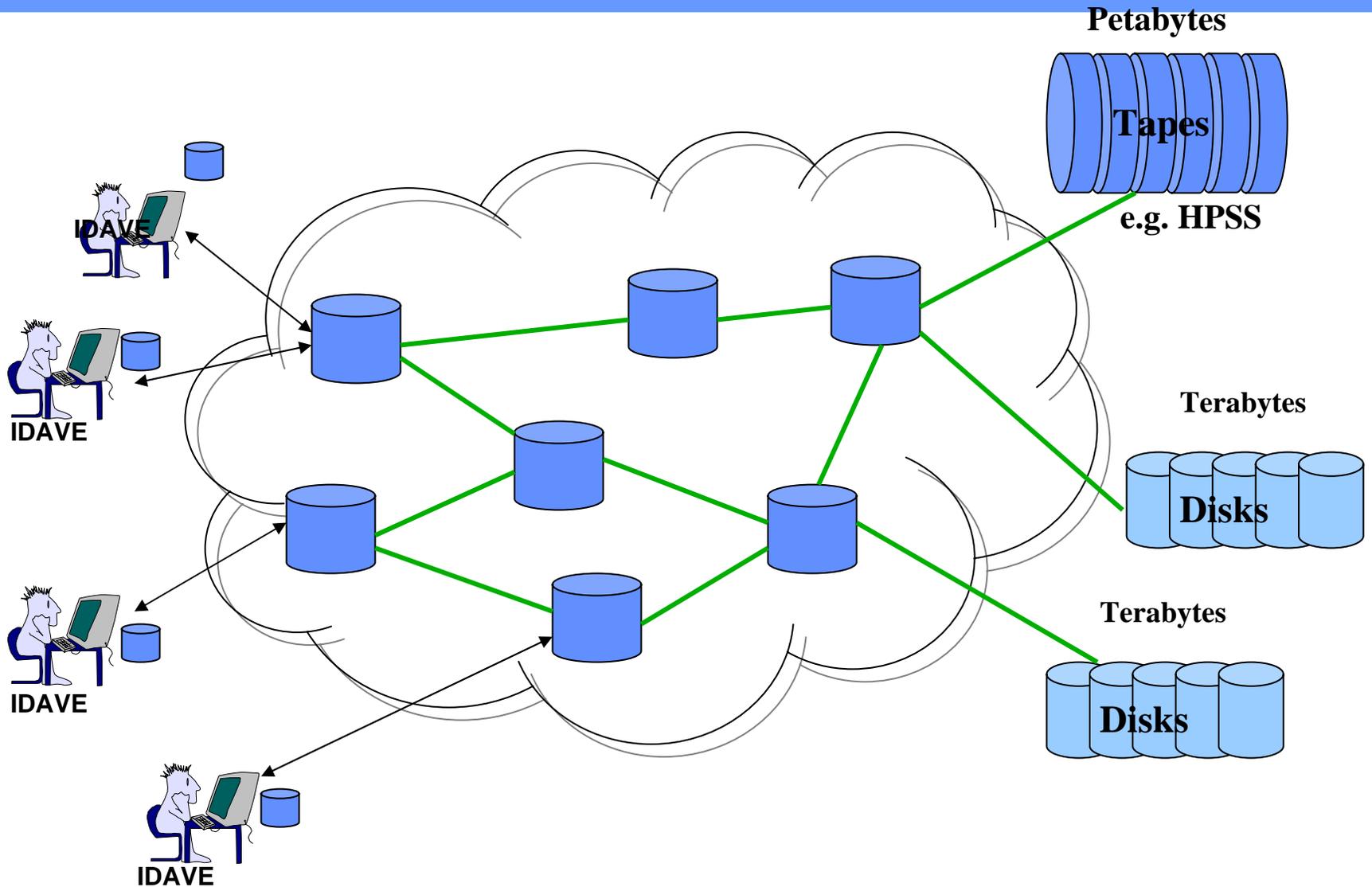
- * Different systems
- * Security procedures
- * File streaming
- * Lifetime of request
 - Garbage collection

❖ Solution

- Storage Resource Managers (SRMs)



Ubiquitous and Transparent Data Sharing



The Scientific Data Management Center

(Center for Enabling Technologies - CET)



FROM DATA TO DISCOVERY

- ❖ PI: Arie Shoshani, LBNL
- ❖ Annual budget: 3.3 Million
- ❖ Established 5 years ago (SciDAC-1)
- ❖ Successfully re-competed for the next 5 years (SciDAC-2)
- ❖ Featured in next issue of SciDAC magazine
- ❖ Laboratories
 - ANL, ORNL, LBNL, LLNL, PNNL
- ❖ Universities
 - NCSU, NWU, SDSC, UCD, Uof Utah,

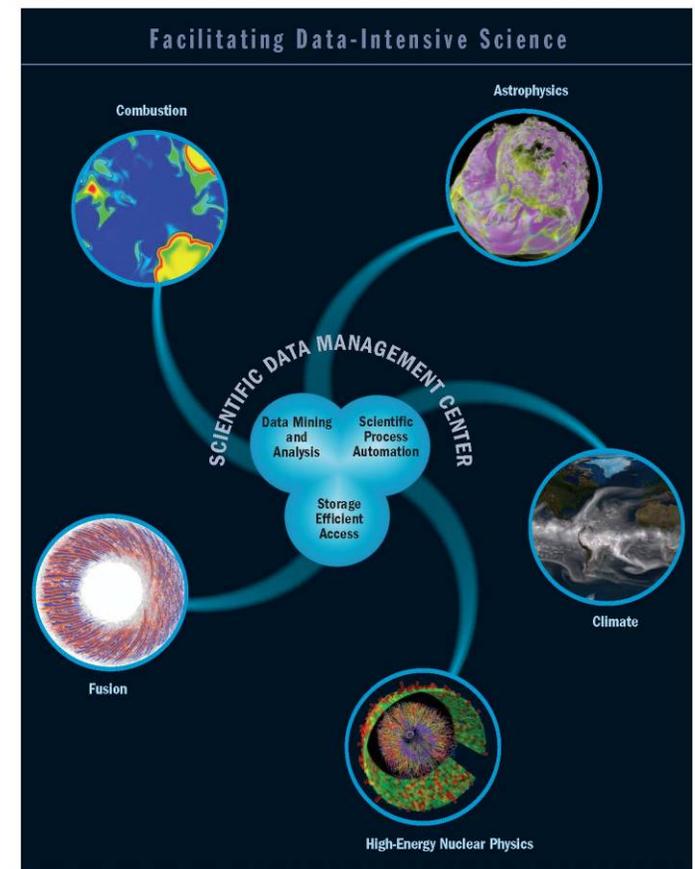
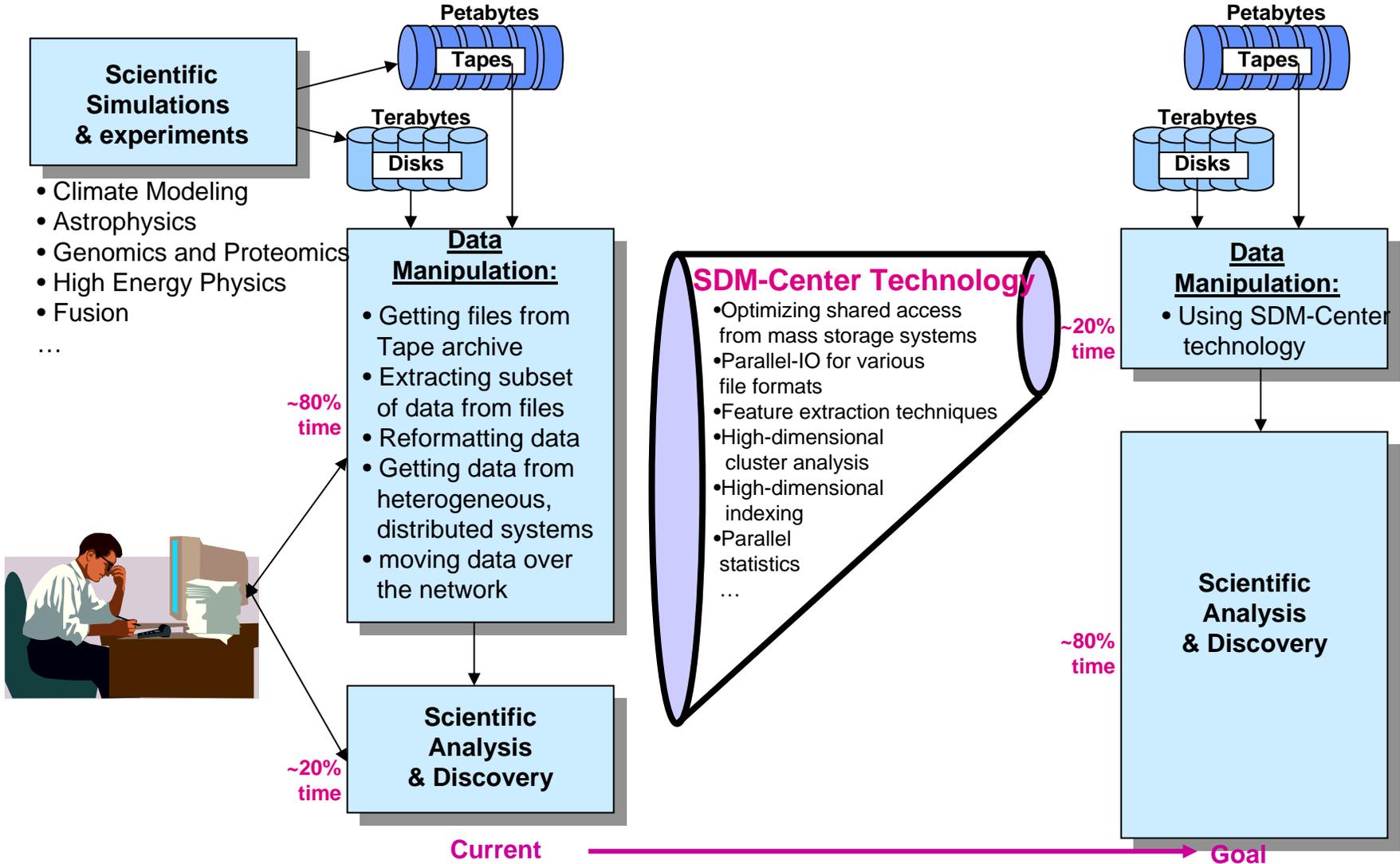
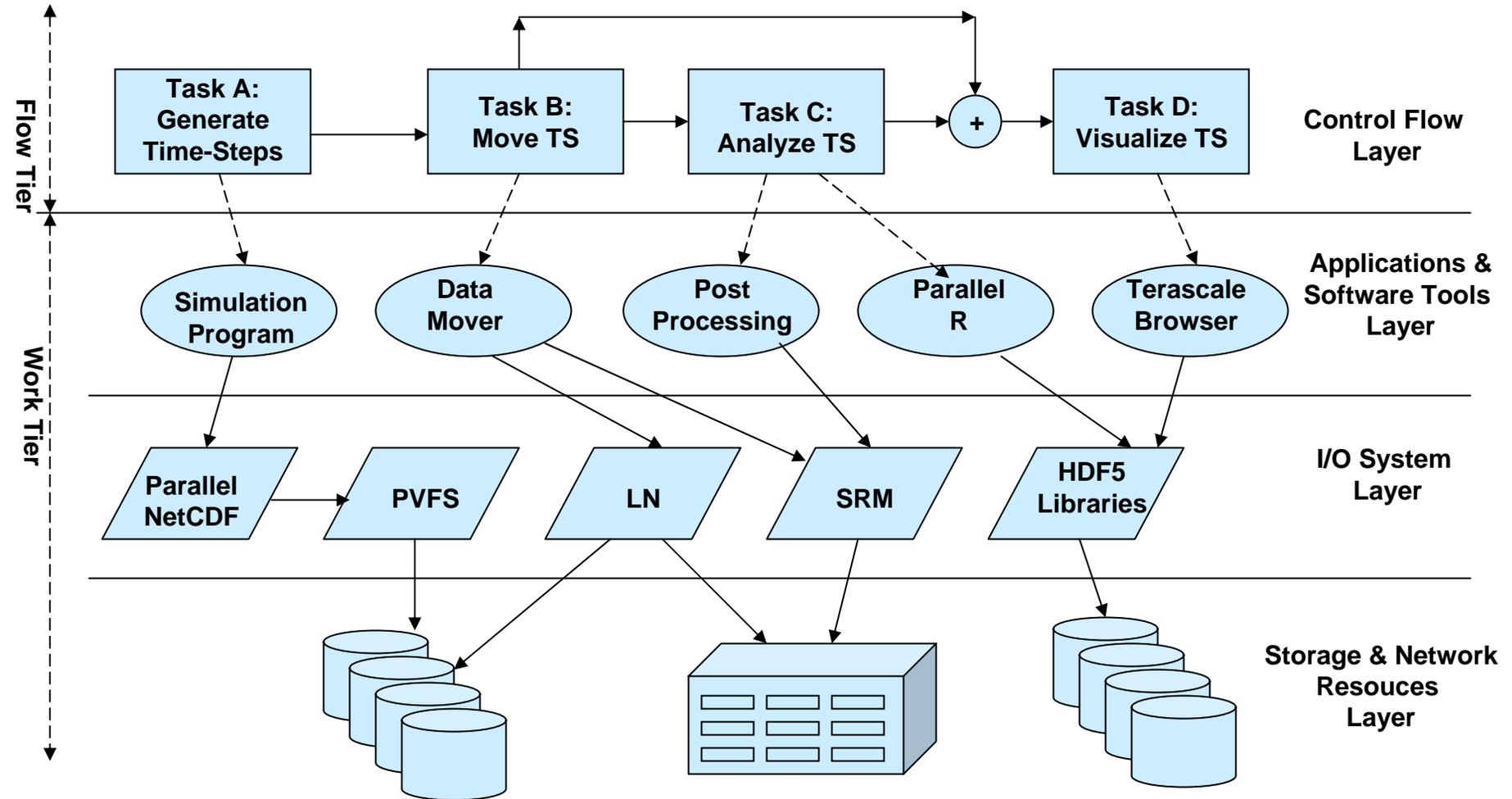


Figure 10. Combustion, Astrophysics, Climate, High-Energy Physics, and Fusion are some of the many research applications that have benefited from the creative technologies offered by the SDM Center

Scientific Data Management Center

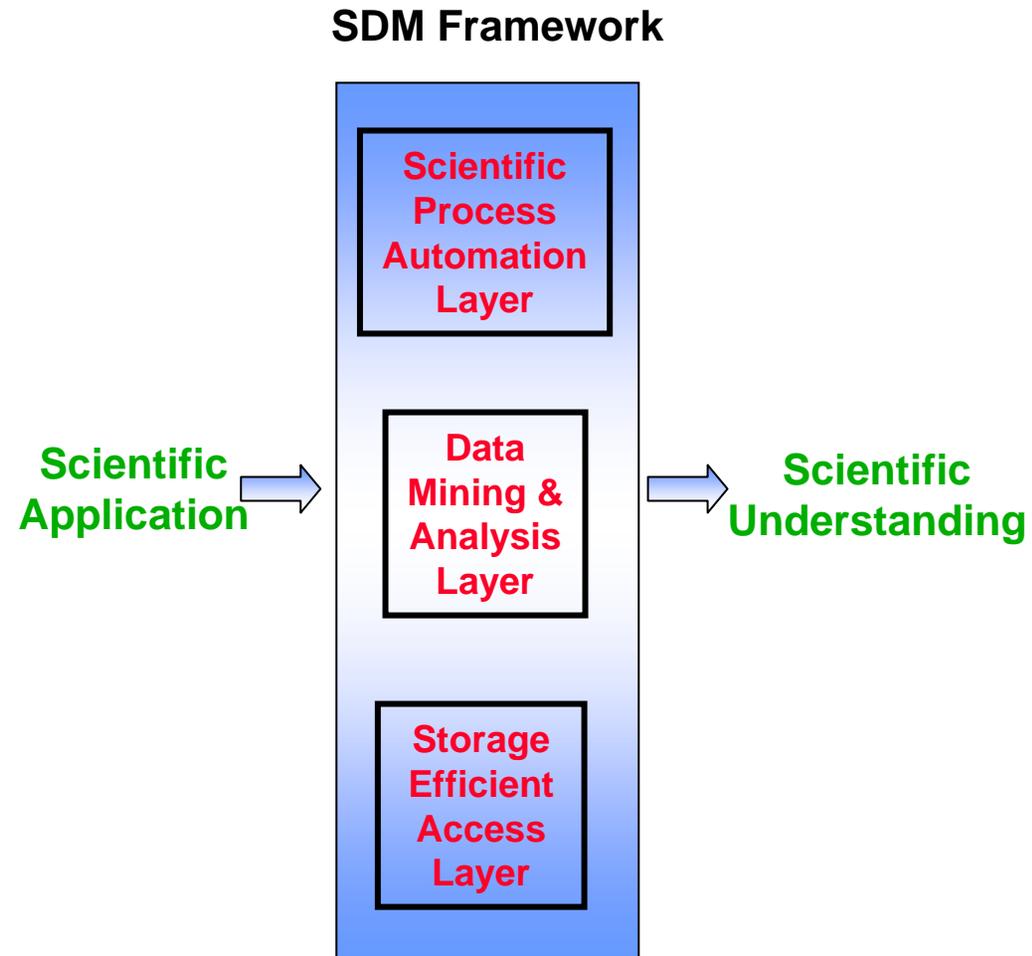


A Typical SDM Scenario



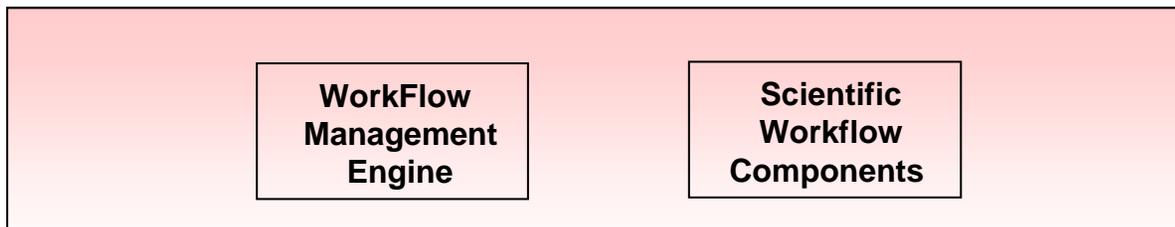
Approach

- Use an integrated framework that:
 - Provides a scientific workflow capability
 - Supports data mining and analysis tools
 - Accelerates storage and access to data
- Simplify data management tasks for the scientist
 - Hide details of underlying parallel and indexing technology
 - Permit assembly of modules using a simple graphical workflow description tool

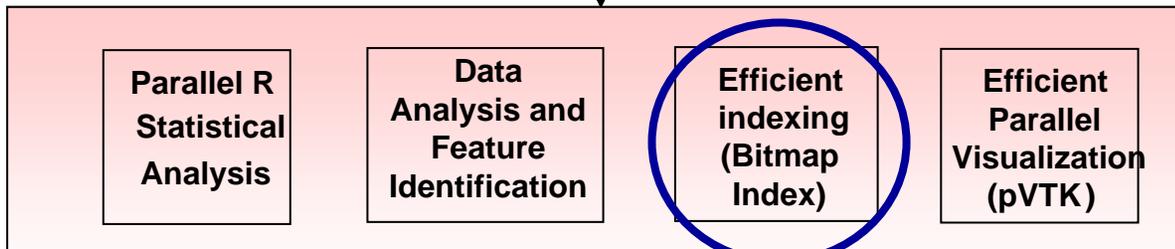


Technology Details by Layer

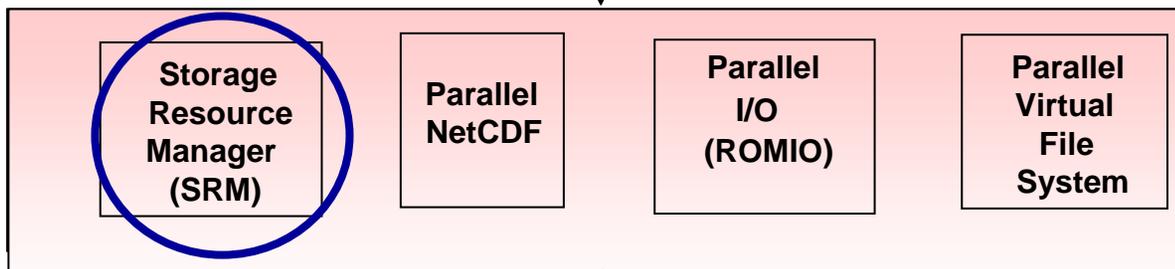
**Scientific
Process
Automation
(SPA)
Layer**



**Data
Mining &
Analysis
(DMA)
Layer**



**Storage
Efficient
Access
(SEA)
Layer**



SDM Technology: Layering of Components

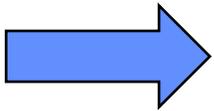
**Scientific Process Automation
(Workflow) Layer**



**Data Mining & Analysis
Layer**



**Storage Efficient Access
Layer**



Hardware, OS, and MSS (HPSS)

Data Generation

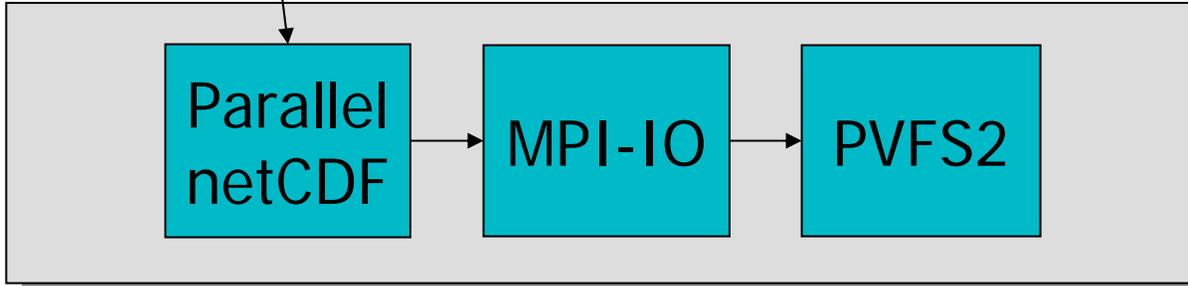
Scientific Process
Automation Layer



Data Mining and
Analysis Layer



Storage Efficient
Access Layer



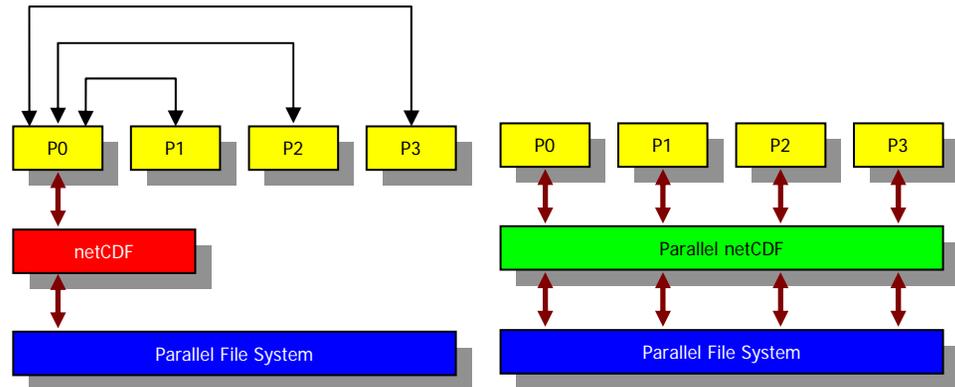
Parallel NetCDF v.s. HDF5 (ANL+NWU)



Parallel Virtual File System: Enhancements and deployment

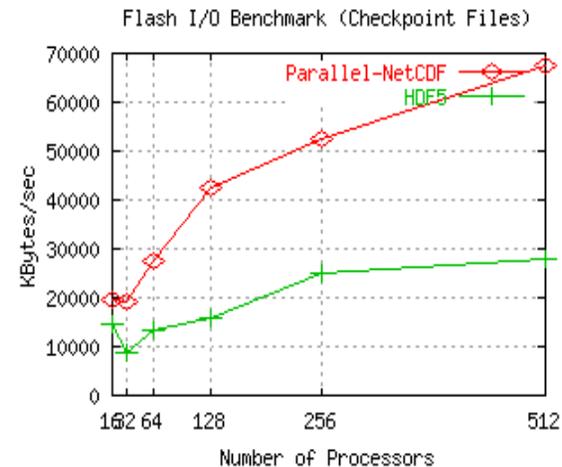
- ❖ Developed Parallel netCDF
 - Enables high performance parallel I/O to netCDF datasets
 - Achieves up to 10 fold performance improvement over HDF5
- ❖ Enhanced ROMIO:
 - Provides MPI access to PVFS2
 - Advanced parallel file system interfaces for more efficient access
- ❖ Developed PVFS2
 - Production use at ANL, Ohio SC, Univ. of Utah HPC center
 - Offered on Dell clusters
 - Being ported to IBM BG/L system
- ❖ Deployed an HPSS Storage Resource Manager (SRM) with PVFS
 - Automatic access of HPSS files to PVFS through MPI-IO library
 - SRM is a middleware component

Interprocess communication



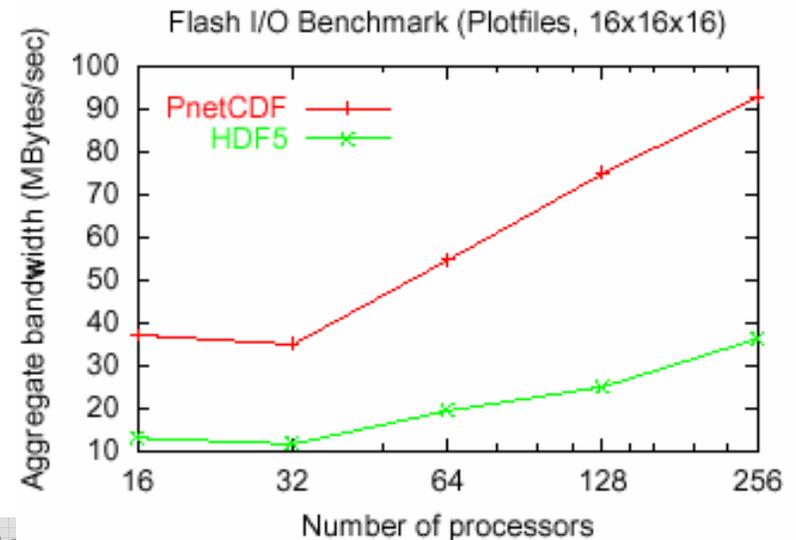
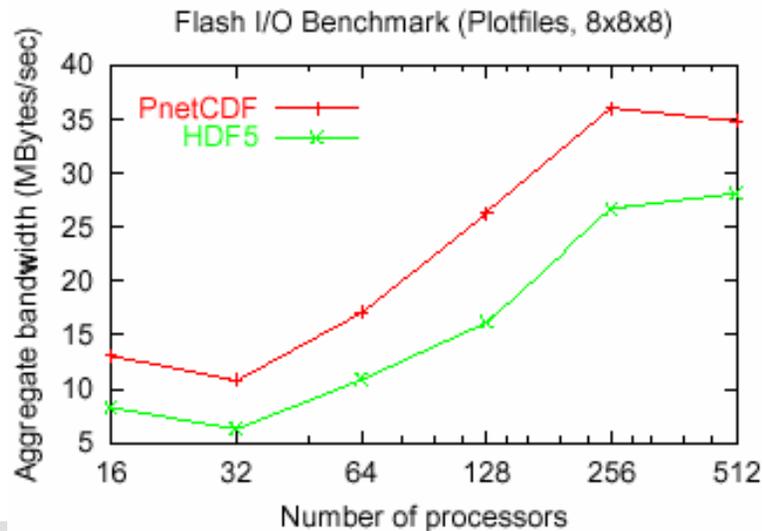
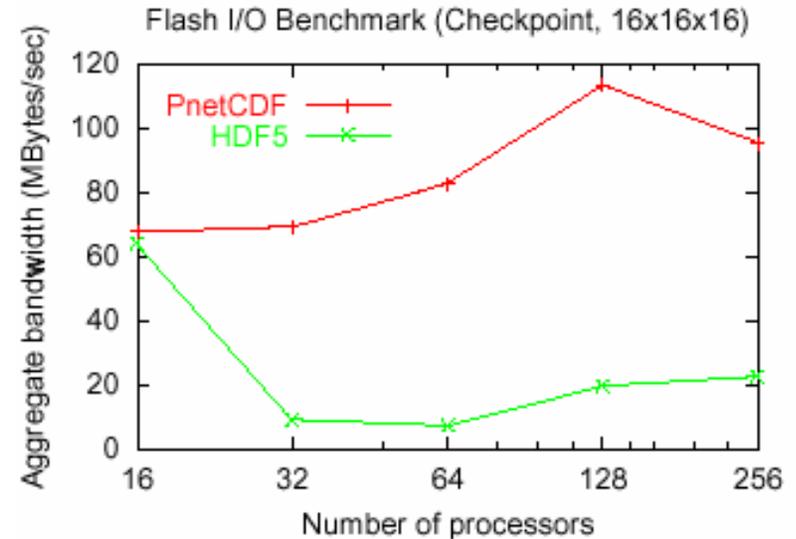
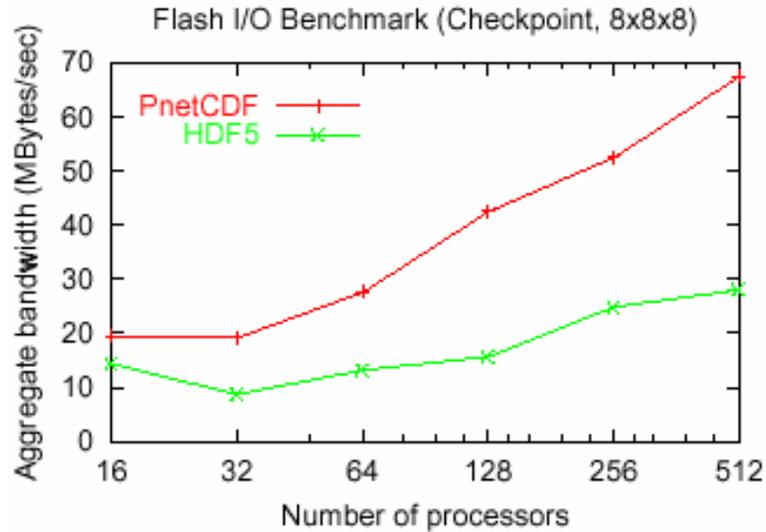
Before

After

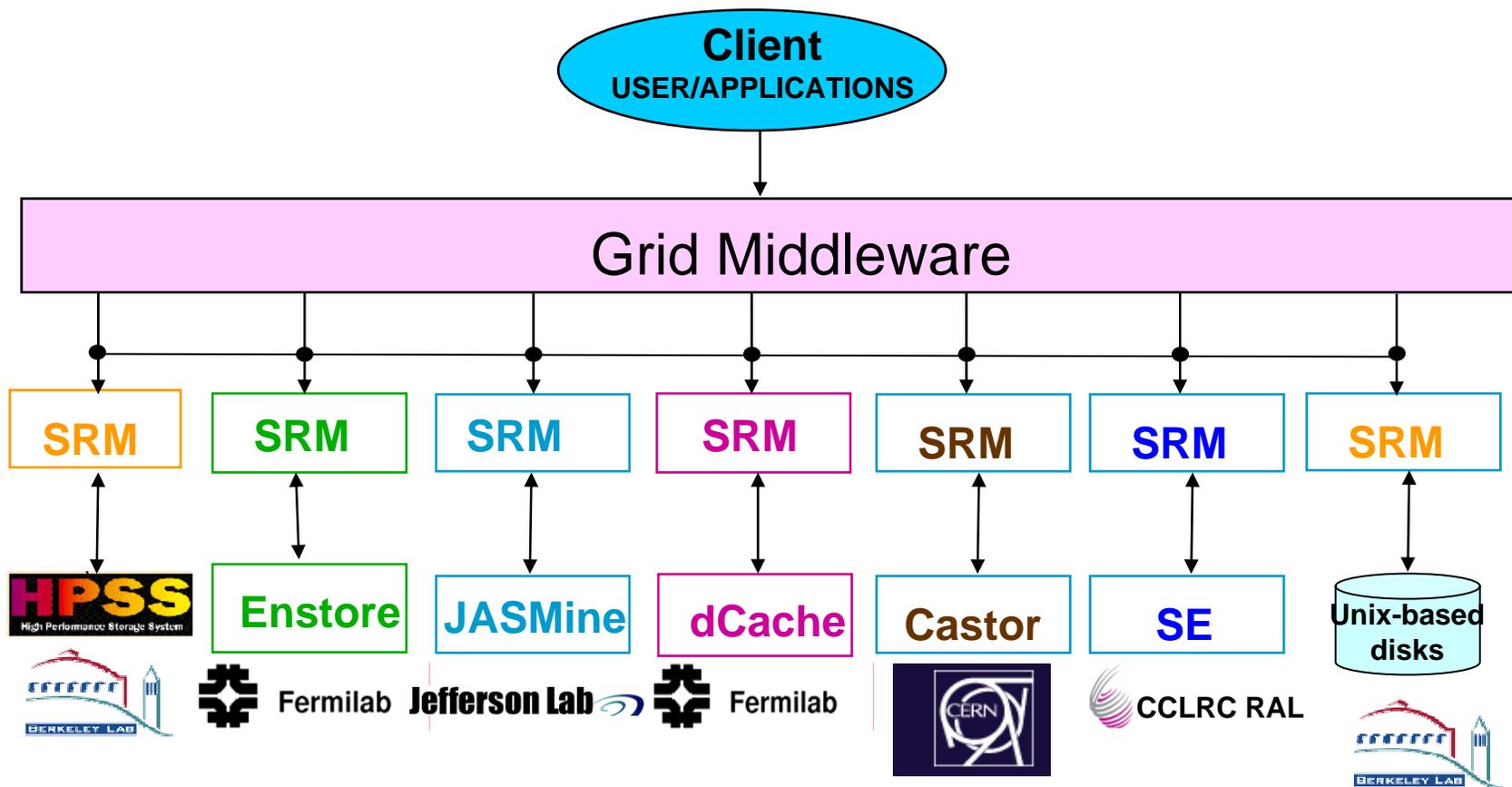


FLASH I/O Benchmark Performance (8x8x8 block sizes)

Parallel netCDF Performance

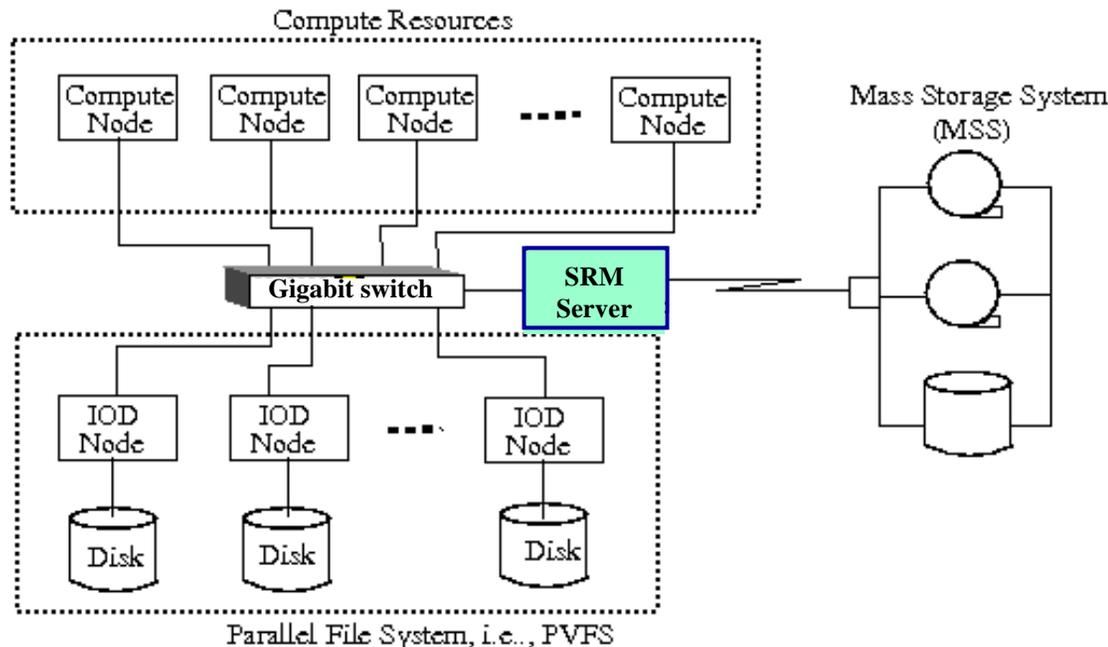


Storage Resource Managers (SRMs)



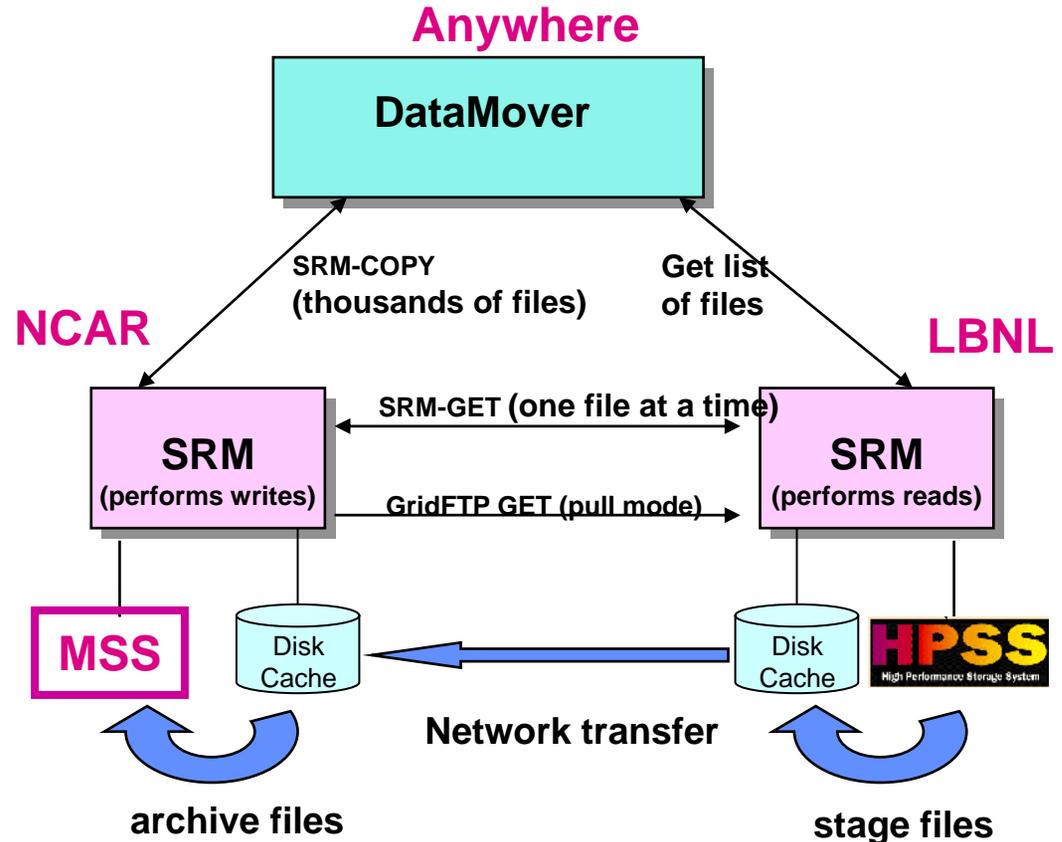
MPI/IO-SRM (LBNL)*

- ❖ Collaboration of LBNL and ANL
 - Access through MPI/IO to files that are on mass storage system
 - Developed MPI/IO-SRM libraries to interact with SRM and HPSS: different hint scenarios (set, bundle, ordered-list)

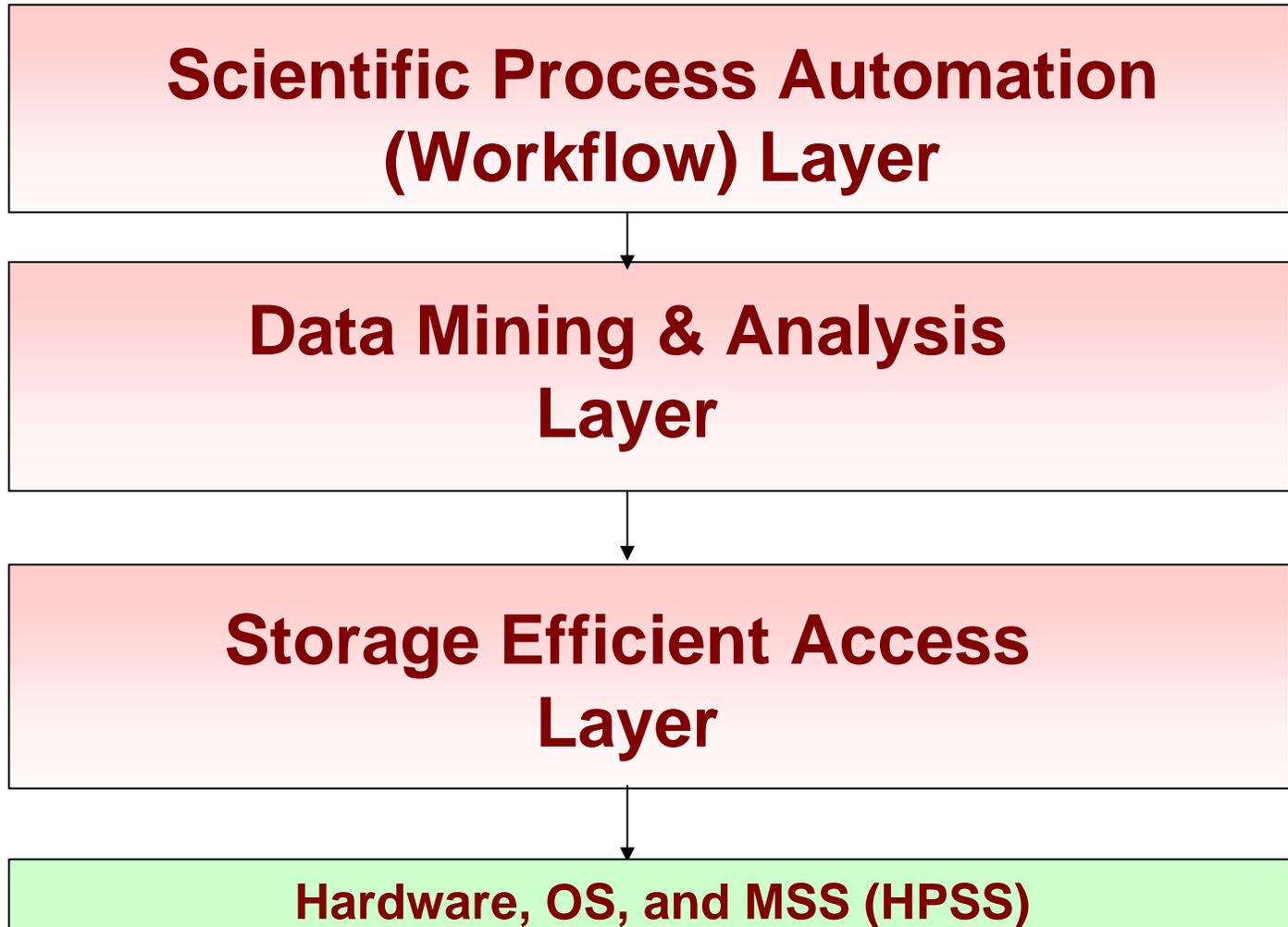


Data Flow Requires Robust Data Movement (LBNL)

- ❖ Problem: move thousands of files robustly
 - Takes many hours
 - Need error recovery
 - Mass storage systems failures
 - Network failures
- ❖ Solution: Use Storage Resource Managers (SRMs)
- ❖ Problem: too slow
- ❖ Solution:
 - in GridFTP
 - Use parallel streams
 - Use large FTP windows
 - Pre-stage files from MSS
 - Use concurrent transfers



Layering of Components



Statistical Computing with R

About R (<http://www.r-project.org/>):

- R is an Open Source (GPL), most widely used programming environment for statistical analysis and graphics; similar to S.
- Provides good support for both users and developers.
- Highly extensible via dynamically loadable add-on packages.
- Originally developed by Robert Gentleman and Ross Ihaka.

```
> library(mva)
> pca <- prcomp(data)
> summary(pca)
> ...
> dyn.load("foo.so")
> .C("foobar")
> dyn.unload("foo.so")
```

```
> library(rpvm)
> .PVM.start.pvmd()
> .PVM.addhosts(...)
> .PVM.config()
```

Towards Enabling Parallel Computing in R:

- **Rmpi** (Hao Yu): R interface to **LAM-MPI**.
- **rpvm** (Na Li and Tony Rossini): R interface to **PVM**; requires knowledge of parallel programming.
- **snow** (Luke Tierney): general API on top of message passing routines to provide high-level (*parallel apply*) commands; mostly demonstrated for *embarrassingly parallel* applications.

	High Level Routines	snow API
parLapply	parallel lapply	
parSapply	parallel sapply	
parApply	parallel apply	
<i>Basic Routines</i>		
clusterExport	export variables to nodes	
clusterCall	call function on each node	
clusterApply	apply function to arguments on nodes	
clusterApplyLB	load balanced clusterApply	
clusterEvalQ	evaluate explicit expression on nodes	
clusterSplit	split vector into pieces for nodes	
<i>Administrative Routines</i>		
makeCluster	create a new cluster of nodes	
stopCluster	shut down a cluster	
clusterSetupSPRNG	initialize random number streams	

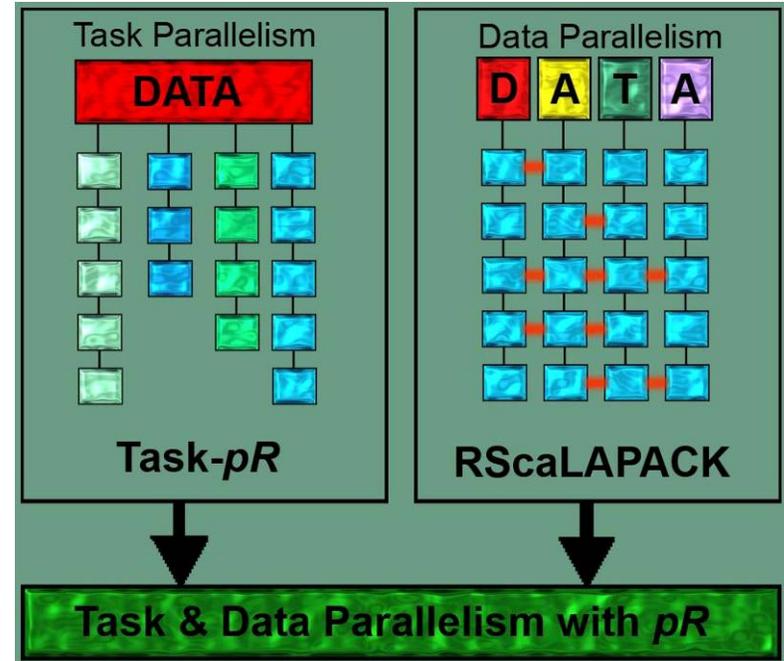
Providing Task and Data Parallelism in *pR*

Task-parallel analyses:

- Likelihood Maximization.
- Re-sampling schemes: Bootstrap, Jackknife, etc.
- Animations
- Markov Chain Monte Carlo (MCMC).
 - Multiple chains.
 - Simulated Tempering: running parallel chains at different “temperature“ to improve mixing.

Data-parallel analyses:

- *k*-means clustering
- Principal Component Analysis (PCA)
- Hierarchical (model-based) clustering
- Distance matrix, histogram, etc. computations



```

.....
fileList<-list.files(pattern="*.nc");
for (i in 1:length(fileList)) {
  matrix [i] ← readNcFile (fileList[i]);
  pca [i] ← prcomp (matrix [i])
}
.....

```

R

```

.....
fileList<-list.files(pattern="*.nc");
PE ( for (i in 1:length(fileList)) {
  matrix [i] ← readNcFile (fileList[i]);
  pca [i] ← sla.prcomp (matrix [i])
} )
.....

```

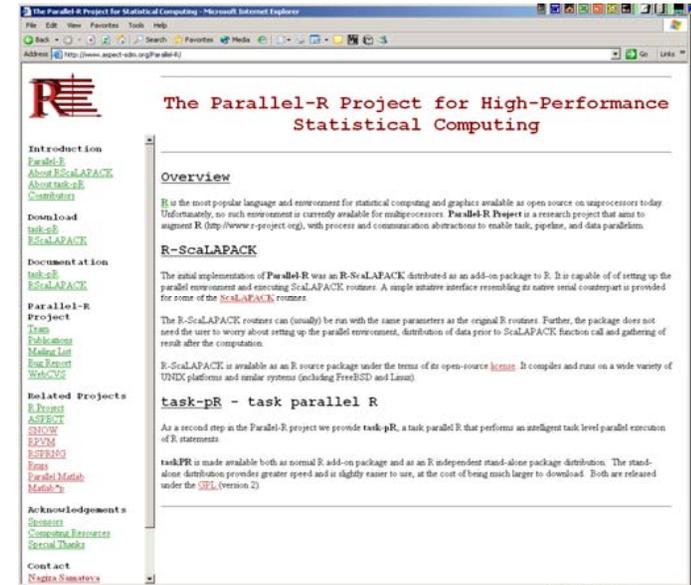
pR

Parallel R (pR) Distribution

Releases History:

- **pR** enables both *data and task parallelism* (includes task-pR and RScalLAPACK) (version 1.8.1)
- **RScalLAPACK** provides **R** interface to **ScaLAPACK** with its scalability in terms of problem size and number of processors using *data parallelism* (release 0.5.1)
- **task-pR** achieves parallelism by performing *out-of-order execution of tasks*. With its intelligent scheduling mechanism it attains significant gain in execution times (release 0.2.7)
- **pMatrix** provides a parallel platform to perform *major matrix operations in parallel* using ScaLAPACK and PBLAS Level II & III routines

<http://www.ASPECT-SDM.org/Parallel-R>

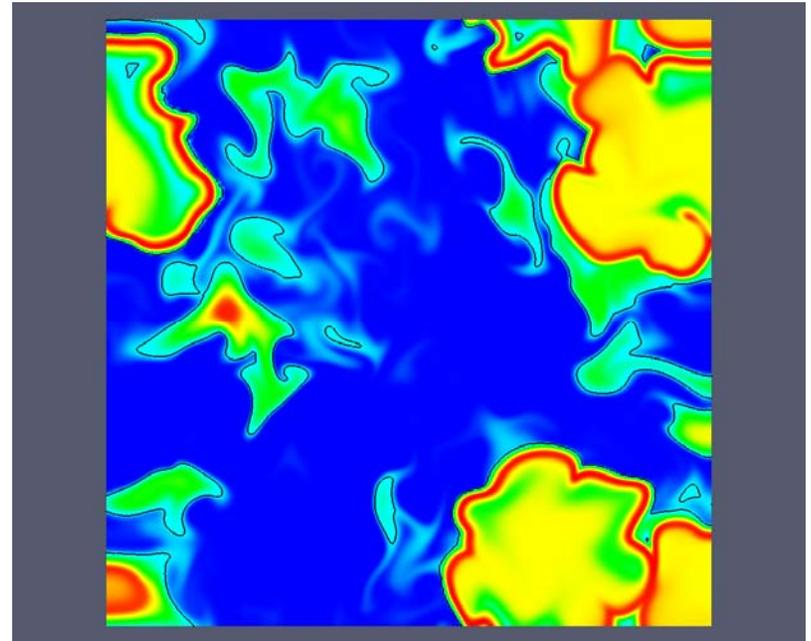


Also: Available for download from **R's CRAN** web site (www.R-Project.org) with **37 mirror sites in 20 countries**

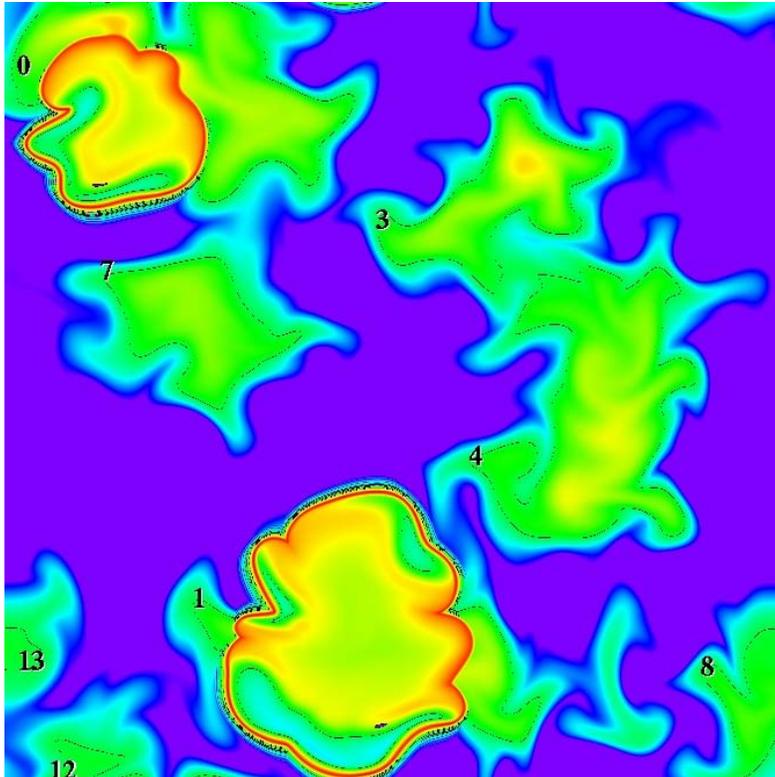
FastBit – Bitmap Indexing Technology (LBNL)



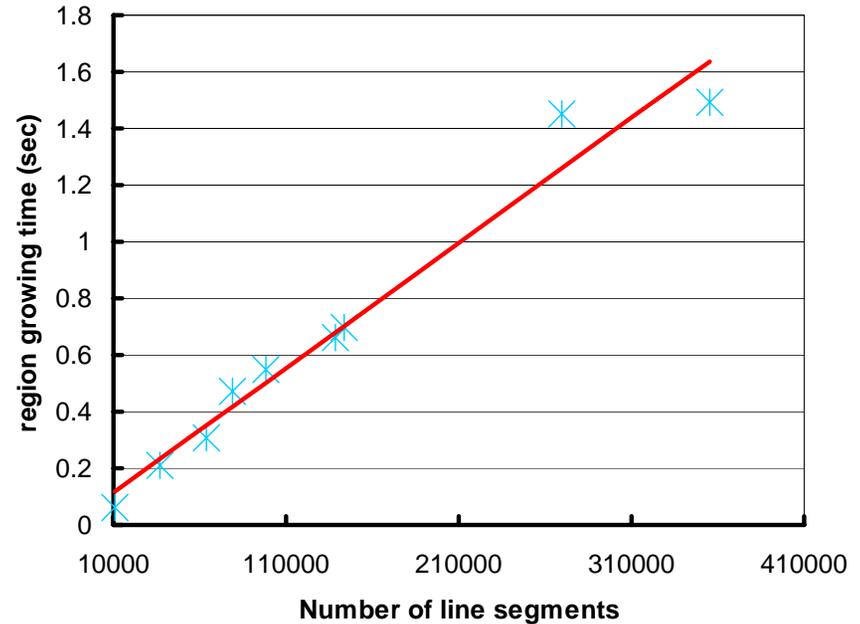
- ❖ Astrophysics or Combustion simulations (Spatio-temporal datasets)
 - 1000x1000x1000 mesh with 10-100s of variables per cell over 1000s of time steps - 10^{13} - 10^{14} data values
- ❖ This is an image of a single variable
- ❖ What's needed is search over multiple variables, such as
- ❖ Temperature > 1000
AND pressure > 10^6
AND $\text{HO}_2 > 10^{-7}$ AND $\text{HO}_2 > 10^{-6}$
- ❖ Combining multiple single-variable indexes - a problem
- ❖ Solution: specialized bitmap indexes (logical operation on resulting bitmaps)
- ❖ Search time proportional to number of hits



FastBit-Based Multi-Attribute Region Finding is Theoretically Optimal



Flame Front discovery
(range conditions for multiple measures)
in a combustion simulation (Sandia)

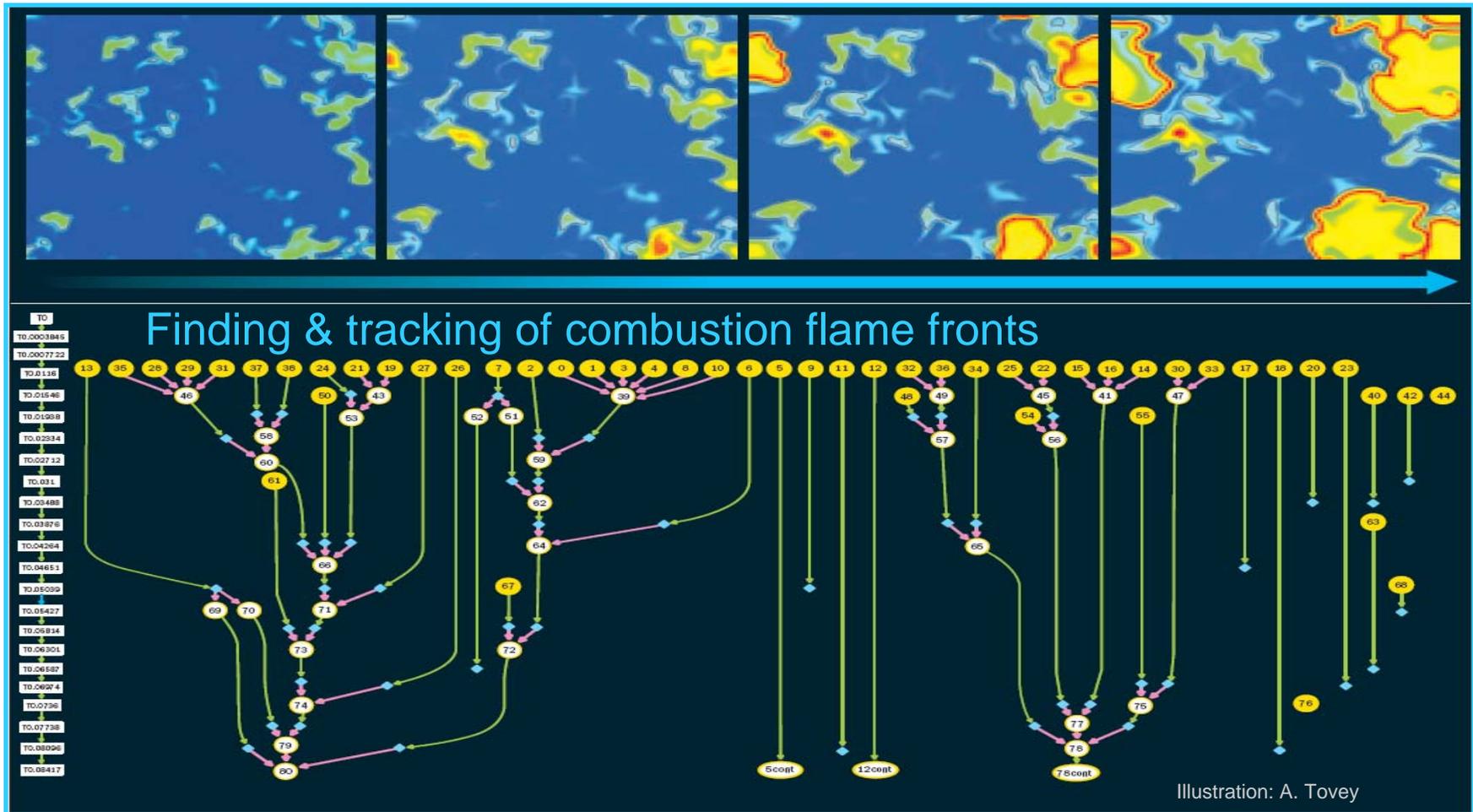


Time required to identify regions in
3D Supernova simulation (LBNL)

On 3D data with over **110 million points**,
region finding takes **less than 2 seconds**

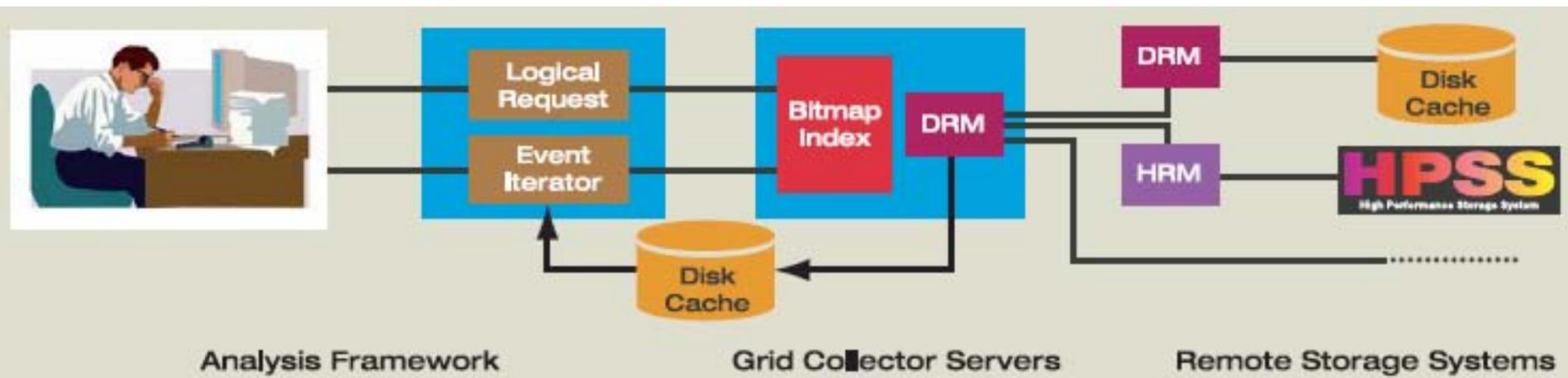
Searching for regions that satisfy particular criteria with FastBit

Plotting progression of flames can be performed in real time



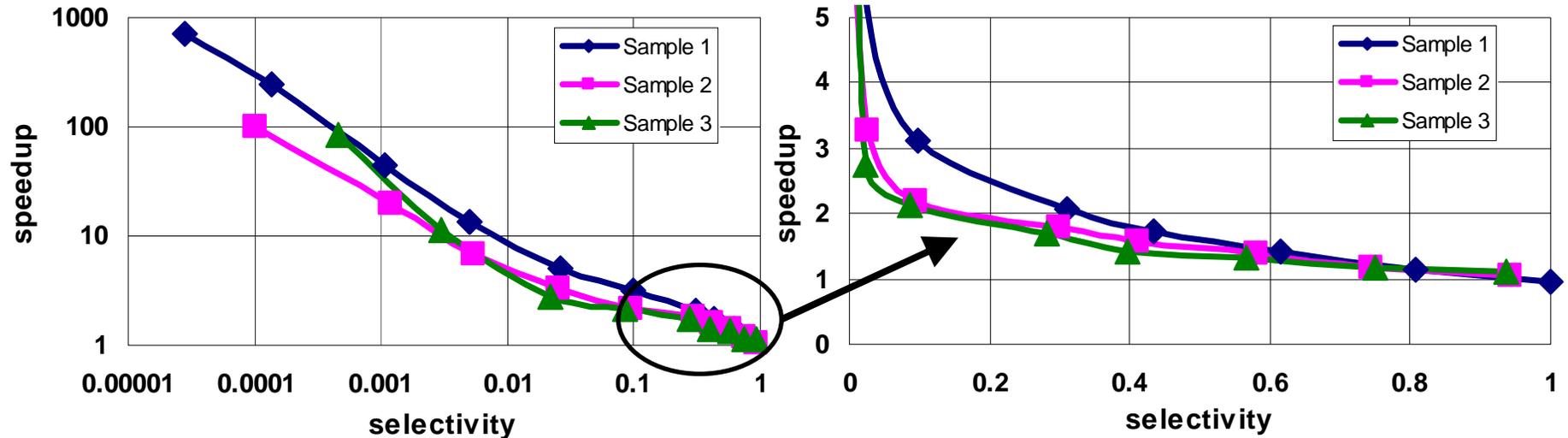
Ollaborators: SNL: Drs. J. Chen, W. Doyle, NCSU: Dr. T. Echecki

Grid Collector



- ❖ Searching over billions of objects with hundreds of descriptors each:
 - search over descriptors of billions of HEP (collision) events
 - Data volume Terabytes-Petabytes
- ❖ Benefits of the Grid Collector:
 - transparent object access
 - Selection of objects based on their attribute values
 - Improvement of analysis system's throughput
 - Interactive analysis of data distributed on the Grid
 - Best Paper Award (ISC'05)

Grid Collector Speeds up Analyses



- ❖ Test machine: 2.8 GHz Xeon, 27 MB/s read speed
- ❖ When searching for rare events, say, selecting one event out of 1000, using GC is 20 to 50 times faster
- ❖ Using GC to read 1/2 of events, speedup > 1.5, 1/10 events, speed up > 2.

Grid Collector Success Stories

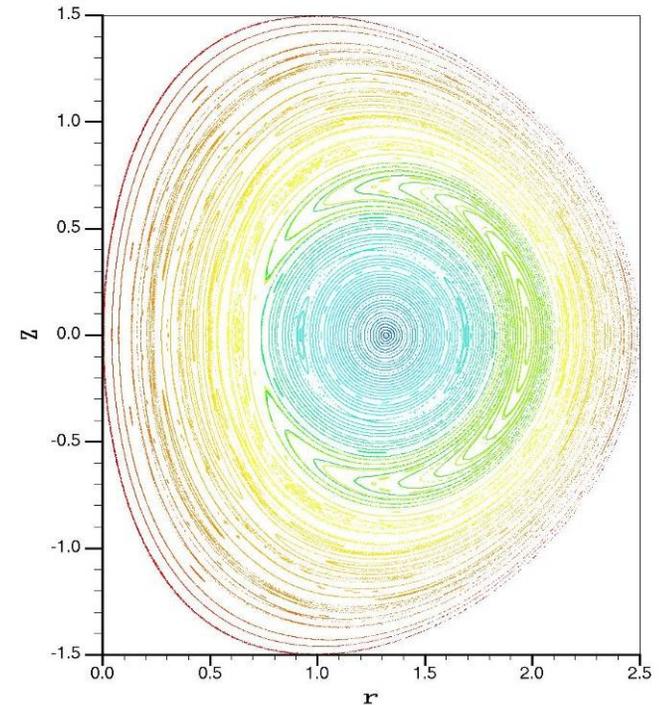


- ❖ Searching for anti- ^3He
- ❖ Lee Barnby, Birmingham
- ❖ Previous studies identified collision events that possibly contain anti- ^3He , need further analysis
- ❖ Searching for strangelets
- ❖ Aihong Tang, BNL
- ❖ Previous studies identified events that behave close to strangelets, need further investigation

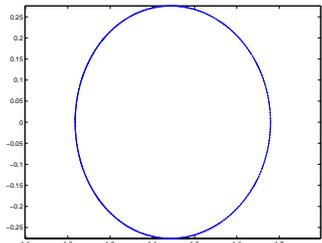
- ❖ Without Grid Collector, one has to retrieve every file from mass storage systems and scan them for the wanted events – may take weeks or months
- ❖ **With Grid Collector, both jobs completed within a day**

Piecewise Polynomial Models for Classification of Puncture (Poincaré) plots

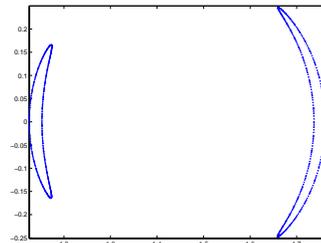
- ❖ Classify each of the nodes:
quasiperiodic, islands,
separatrix
- ❖ Connections between the nodes
- ❖ Want accurate and robust
classification, valid when few
points in each node



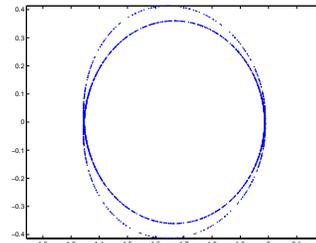
National Compact
Stellarator Experiment



Quasiperiodic



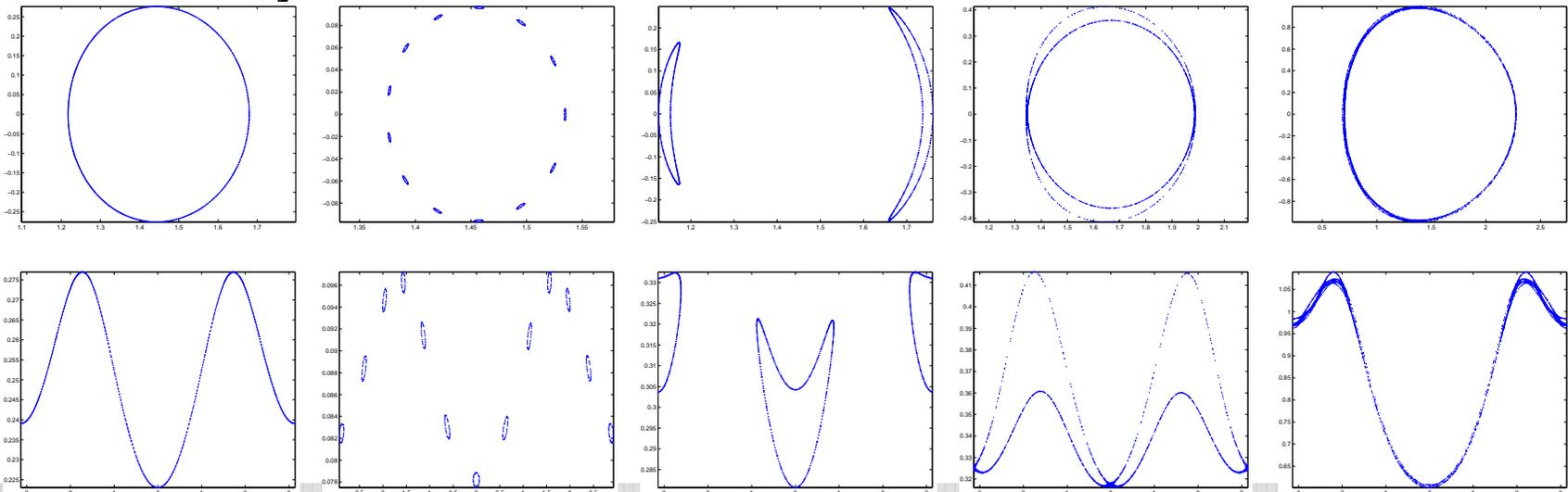
Islands



Separatrix

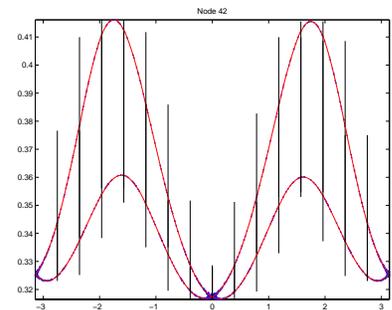
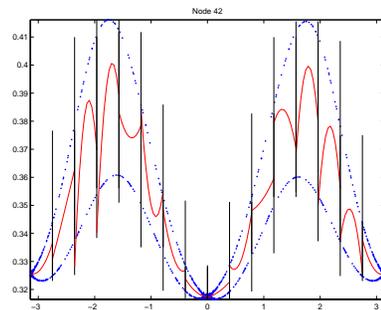
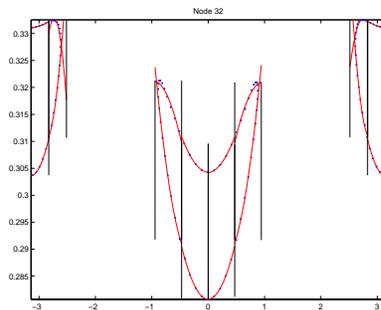
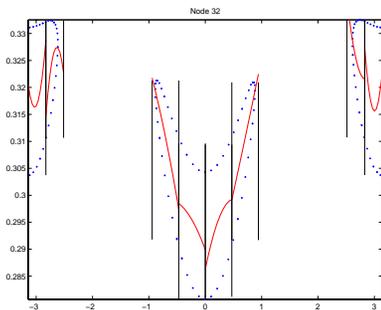
Polar Coordinates

- ❖ Transform the (x,y) data to Polar coordinates (r,θ) .
- ❖ Advantages of polar coordinates:
 - Radial exaggeration reveals some features that are hard to see otherwise.
 - Automatically restricts analysis to radial band with data, ignoring inside and outside.
 - Easy to handle rotational invariance.



Piecewise Polynomial Fitting: Computing polynomials

- ❖ In each interval, compute the polynomial coefficients to fit 1 polynomial to the data.
- ❖ If the error is high, split the data into an upper and lower group. Fit 2 polynomials to the data, one to each group.

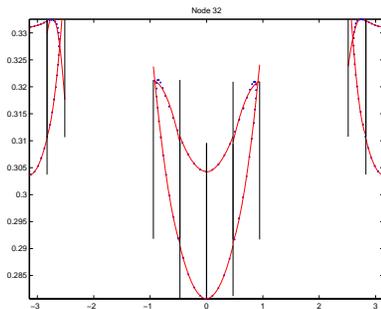


Blue: data. Red: polynomials. Black: interval boundaries.

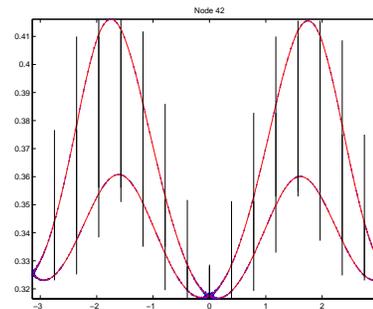
Classification

- ❖ The number of polynomials needed to fit the data and the number of gaps gives the information needed to classify the node:

Gaps	Number of polynomials	
	one	two
Zero	Quasiperiodic	Separatrix
> Zero		Islands

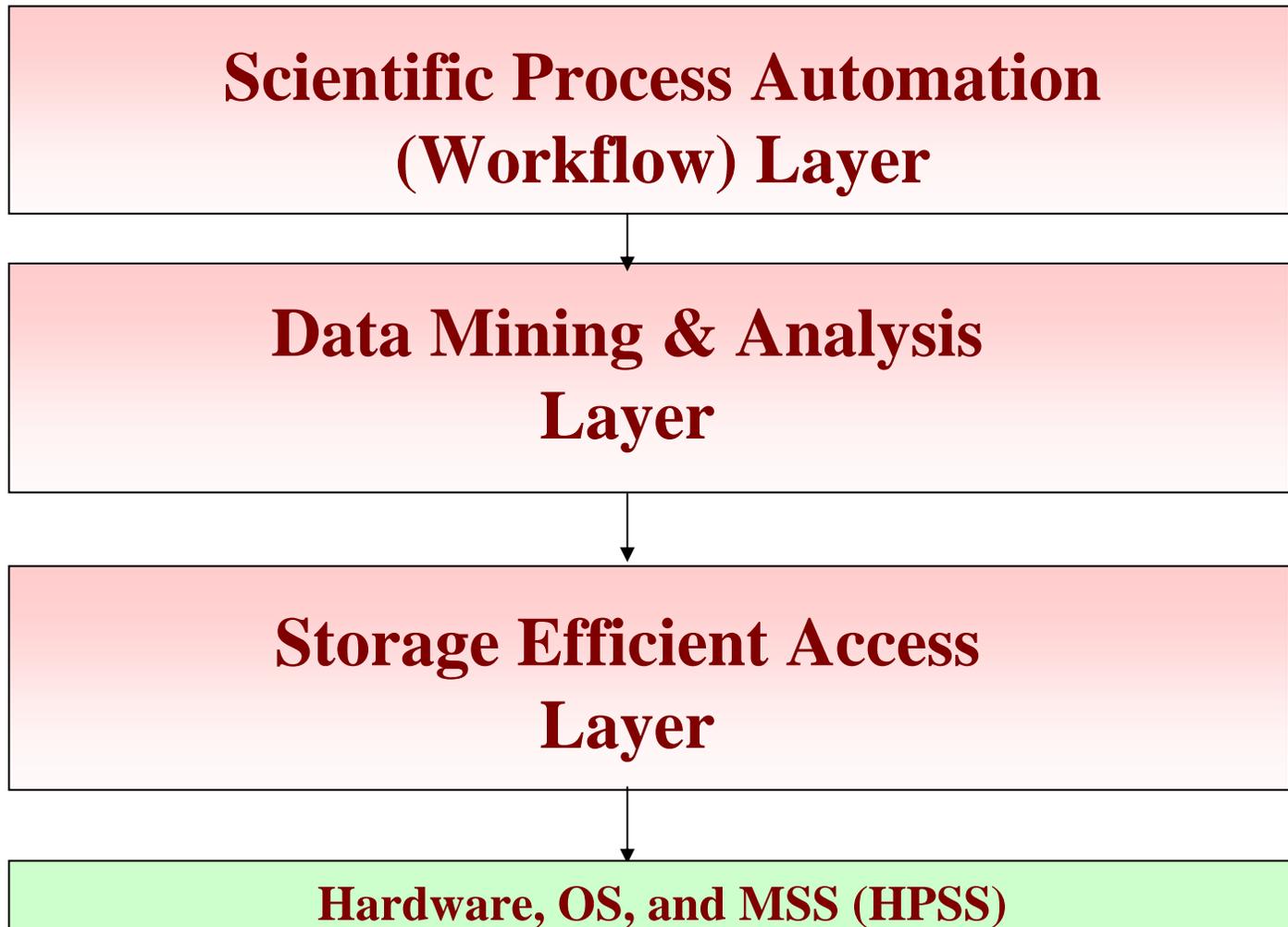
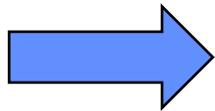


2 Polynomials
2 Gaps
→ Islands

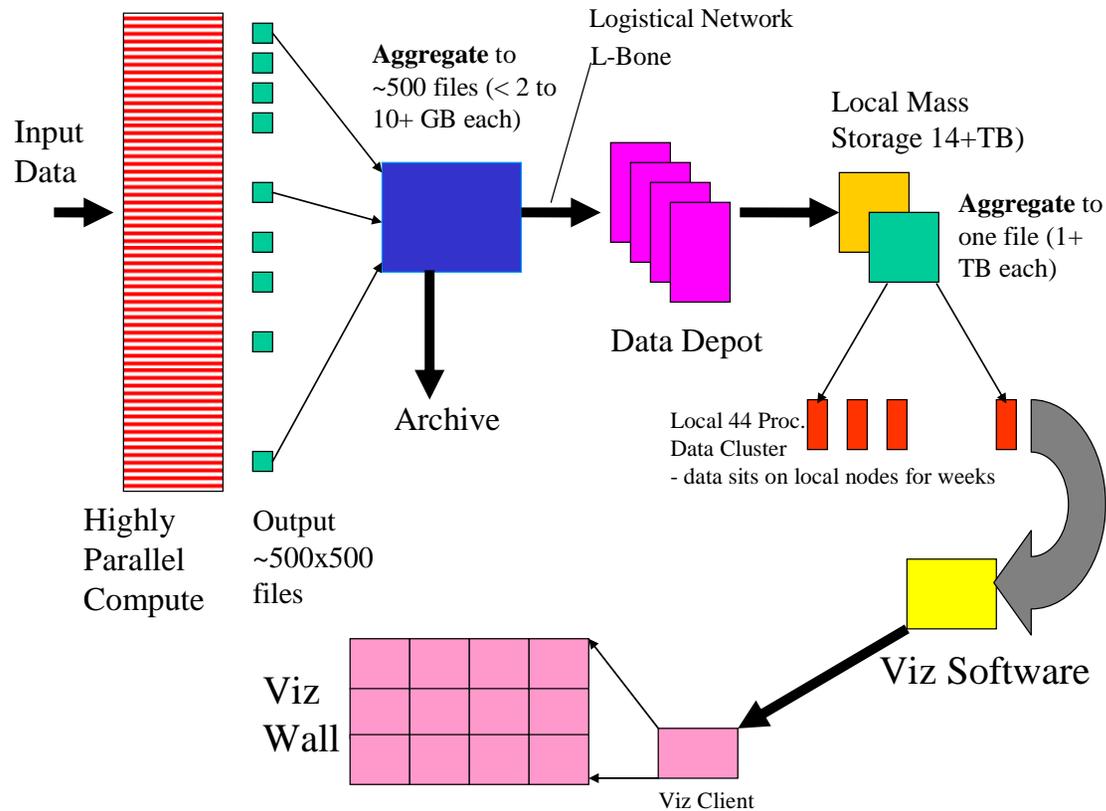


2 Polynomials
0 Gaps
→ Separatrix

Layering of Components

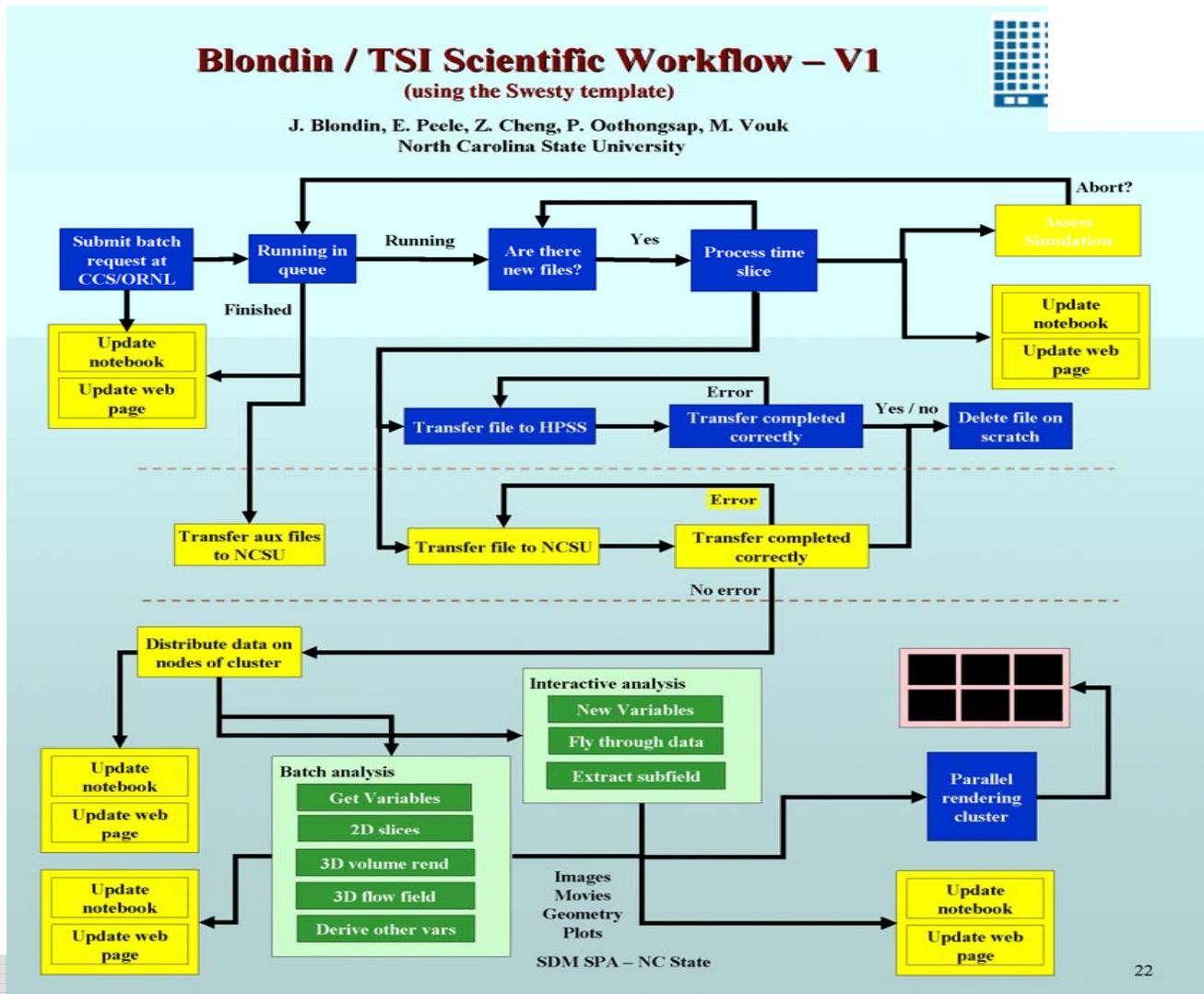


Example Data Flow in TSI



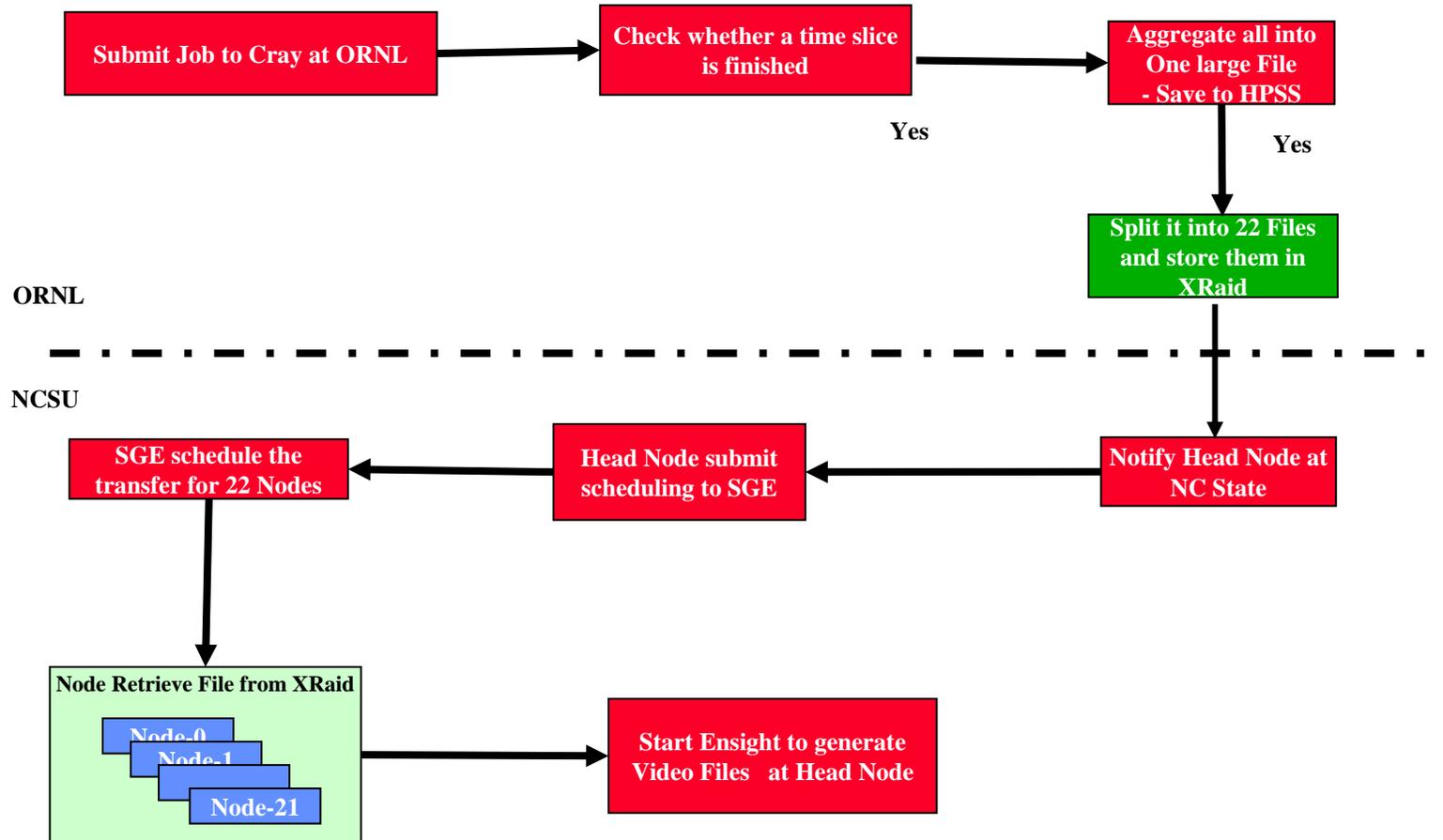
Original TSI Workflow Example with John Blondin, NCSU

Automate data generation, transfer and visualization of a large-scale simulation at ORNL



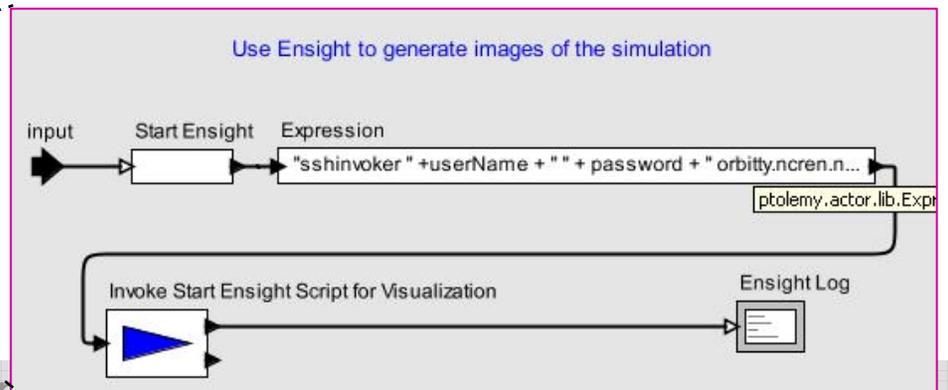
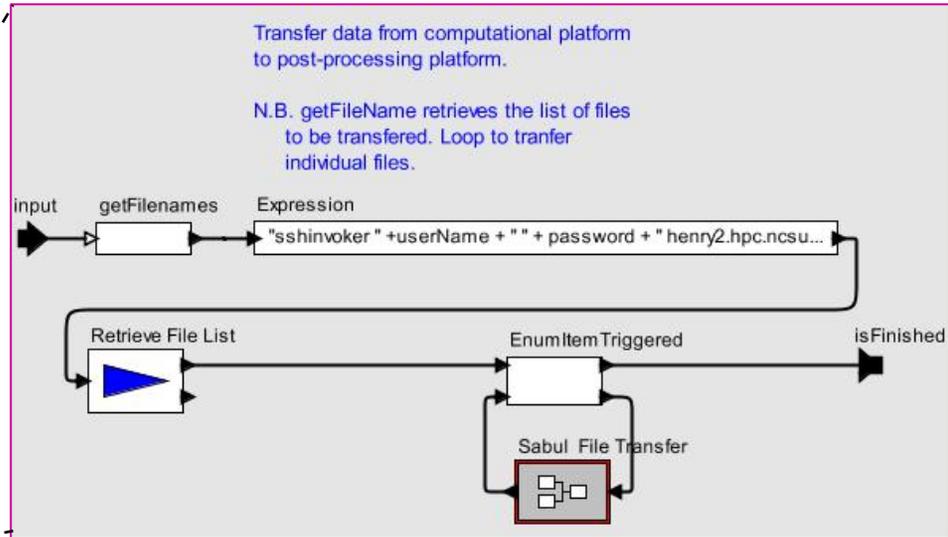
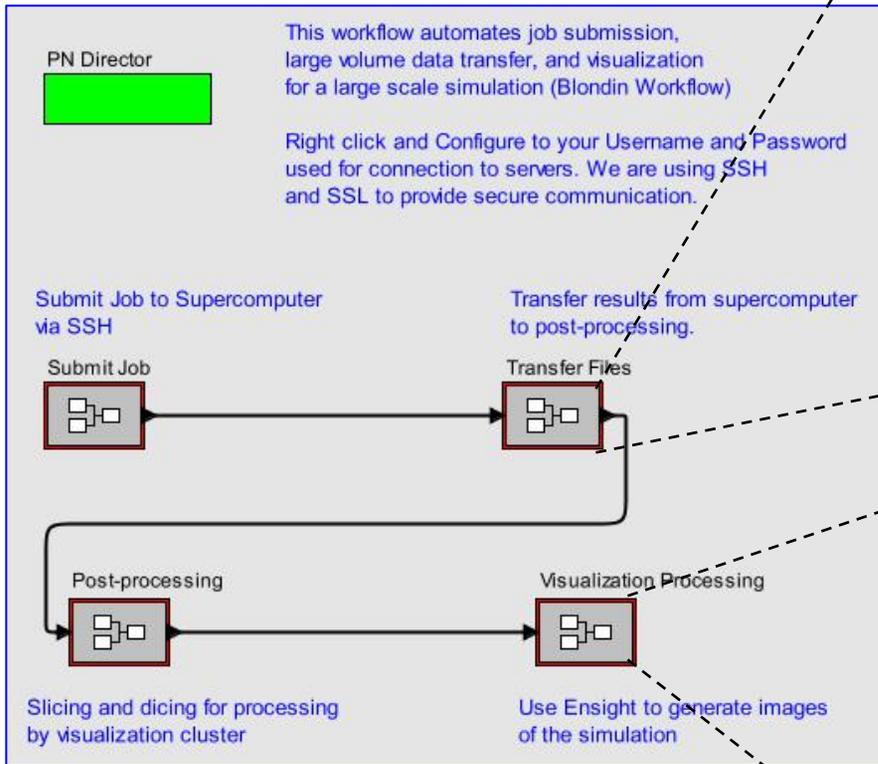
Top level TSI Workflow

Automate data generation, transfer and visualization of a large-scale simulation at ORNL

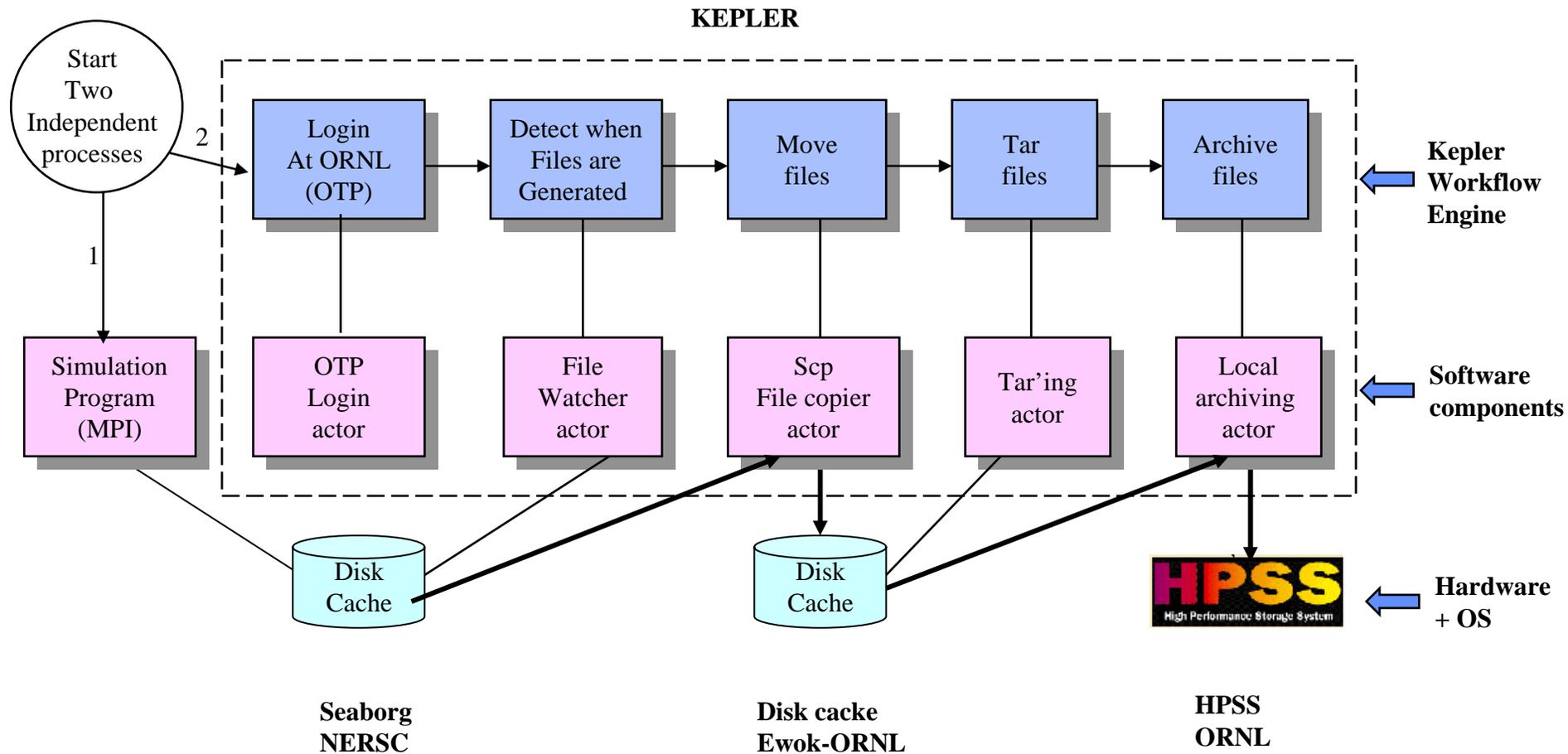


Using the Scientific Workflow Tool (Kepler) Emphasizing Dataflow (SDSC, NCSU, LLNL)

Automate data generation, transfer and visualization of a large-scale simulation at ORNL



New actors in CPES workflow to overcome problems



Re-applying Technology

SDM technology, developed for one application, can be effectively targeted at many other applications ...

Technology	Initial Application	New Applications
Parallel NetCDF	Astrophysics	Climate
Parallel VTK	Astrophysics	Climate
Compressed bitmaps	HENP	Combustion, Astrophysics
Storage Resource Managers	HENP	Astrophysics
Feature Selection	Climate	Fusion (exp. & simulation)
Scientific Workflow	Biology	Astrophysics

Broad Impact of the SDM Center...

Astrophysics:

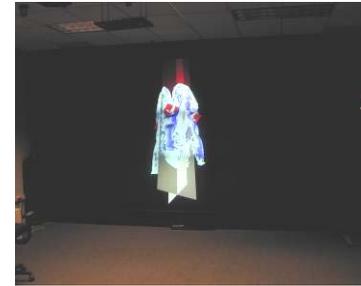
High speed storage technology, parallel NetCDF, parallel VTK, and ASPECT integration software used for Terascale Supernova Initiative (TSI) and FLASH simulations

Tony Mezzacappa – ORNL, Mike Zingale – U of Chicago, Mike Papka – ANL

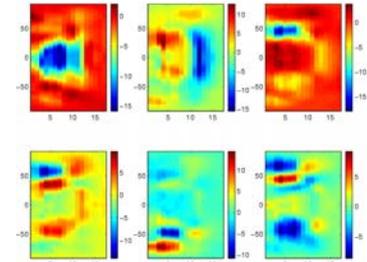
Scientific Workflow

John Blondin – NCSU

Doug Swesty, Eric Myra – Stony Brook



ASCI FLASH – parallel NetCDF

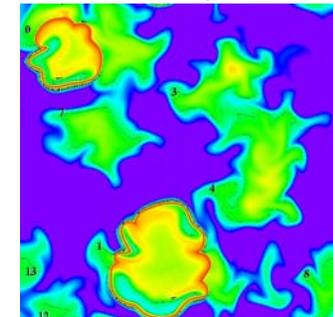


Dimensionality reduction

Climate:

High speed storage technology, Parallel NetCDF, and ICA technology used for Climate Modeling projects

Ben Santer – LLNL, John Drake – ORNL, John Michalakes – NCAR



Region growing

Combustion:

Compressed Bitmap Indexing used for fast generation of flame regions and tracking their progress over time

Wendy Koezler, Jacqueline Chen – Sandia Lab

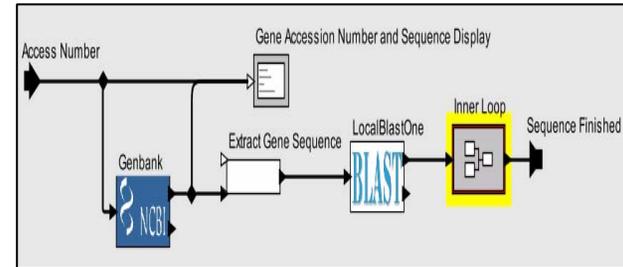
Broad Impact (cont.)*

Biology:

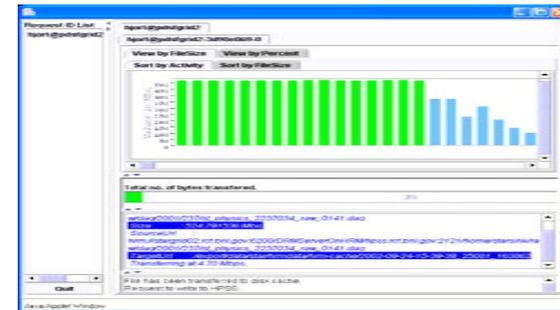
Kepler workflow system and web-wrapping technology used for executing complex highly repetitive workflow tasks for processing microarray data

Matt Coleman – LLNL

Doug Swesty – U of Brookhaven



Building a scientific workflow



Dynamic monitoring of HPSS file transfers

High Energy Physics:

Compressed Bitmap Indexing and Storage Resource Managers used for locating desired subsets of data (events) and automatically retrieving data from HPSS

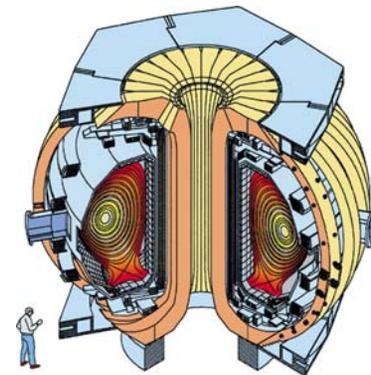
Doug Olson - LBNL, Eric Hjort – LBNL, Jerome Lauret - BNL

Fusion:

A combination of PCA and ICA technology used to identify the key parameters that are relevant to the presence of edge harmonic oscillations in a Tokamak

Keith Burrell - General Atomics

Scott Klasky - PPPL



Identifying key parameters for the DIII-D Tokamak

- ❖ Scientific Data Management poses many challenges
 - Large volume
 - Indexing, transparent file access, storage management, ...
 - Complexity of data model
 - Mesh data, spatio-temporal, sequence data, ...
 - Distributed data and file management
 - Keep track of data location and replication, space management, ...
 - Metadata
 - automatic capture, data entry, provenance (history)
- ❖ Need to support various activities
 - Research, prototype, development, deployment, scaling, maintenance
- ❖ The Data Management Group
 - Focuses research program on problems driven by science applications
 - Successful research prototypes hardened and deployed through collaborative projects, including the SDM center
 - Support for deployed software is an important part of our activities
 - Productivity: papers publication, software development and deployment and support, international service activities (standards, conferences, reviews)

Data Management Requirements for Scientific Applications



Fundamental Technology Areas

- ❖ From the report from the DOE Office of Science Data-Management Workshops (March – May 2004)
 - Efficient access and queries, data integration
 - Distributed data management, data movement, networks
 - Storage and caching
 - Data analysis, visualization, and integrated environments
 - Metadata, data description, logical organization
 - Workflow, data flow, data transformation

Problem: Unique requirements of scientific WFs

- ❖ **Moving large volumes between modules**
 - Tightly-coupled efficient data movement
 - Wide-area loosely coupled data movement
- ❖ **Specification of granularity-based iteration**
 - e.g. In spatial-temporal simulations – a time step is a “granule”
- ❖ **Support for data transformation**
 - complex data types (including file formats, e.g. netCDF, HDF)
- ❖ **Dynamic control of workflow by user**
 - Dynamic user examination of results

Storage Resource Management

Alex Sim
Junmin Gu
Vijaya Natarajan
Alex Romosan
(Arie Shoshani: Design)

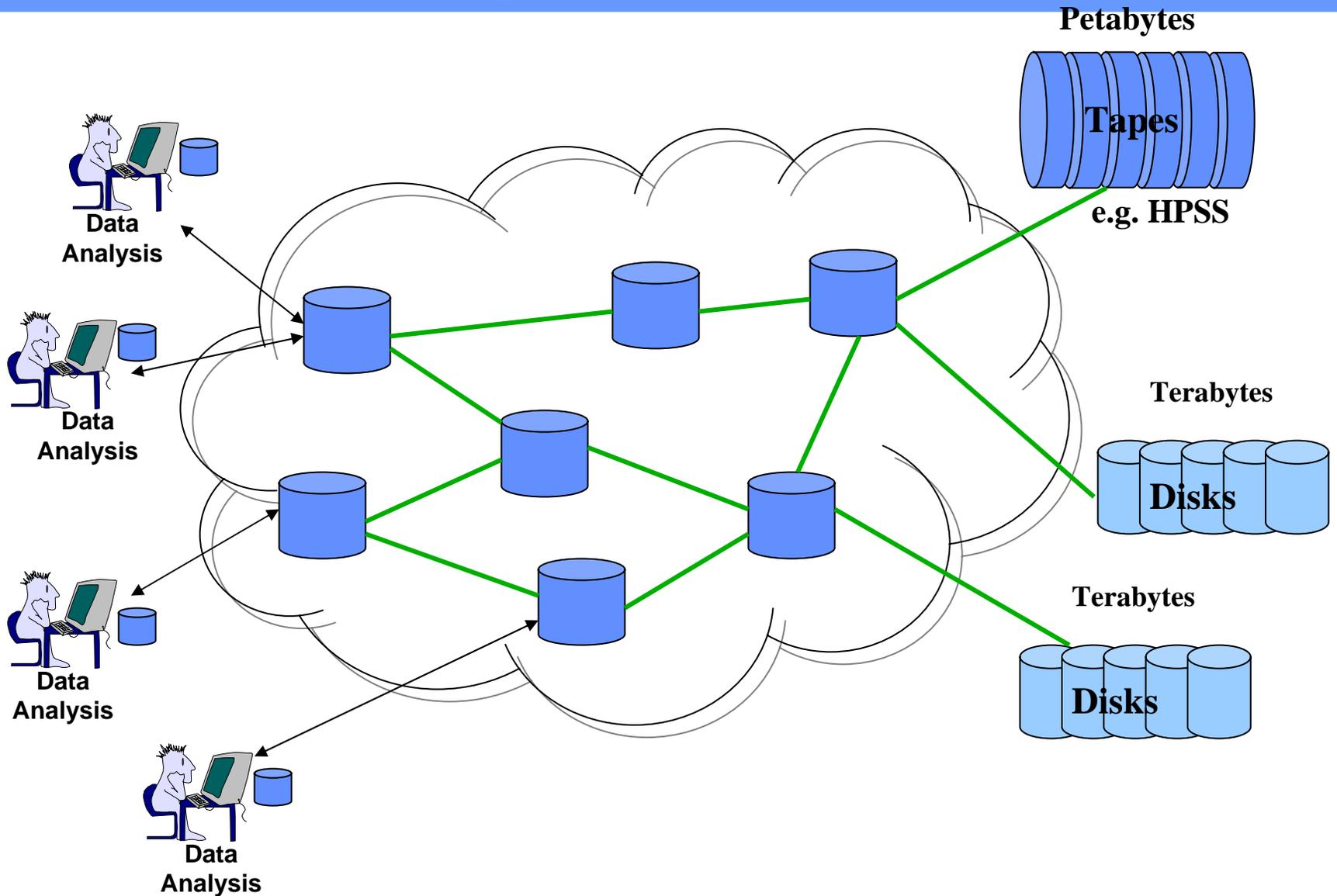
What is SRM?

- ❖ Storage Resource Managers (SRM) are middleware components whose function is to provide
 - Dynamic space allocation
 - Dynamic file management in space
- ❖ For shared storage components on the WAN

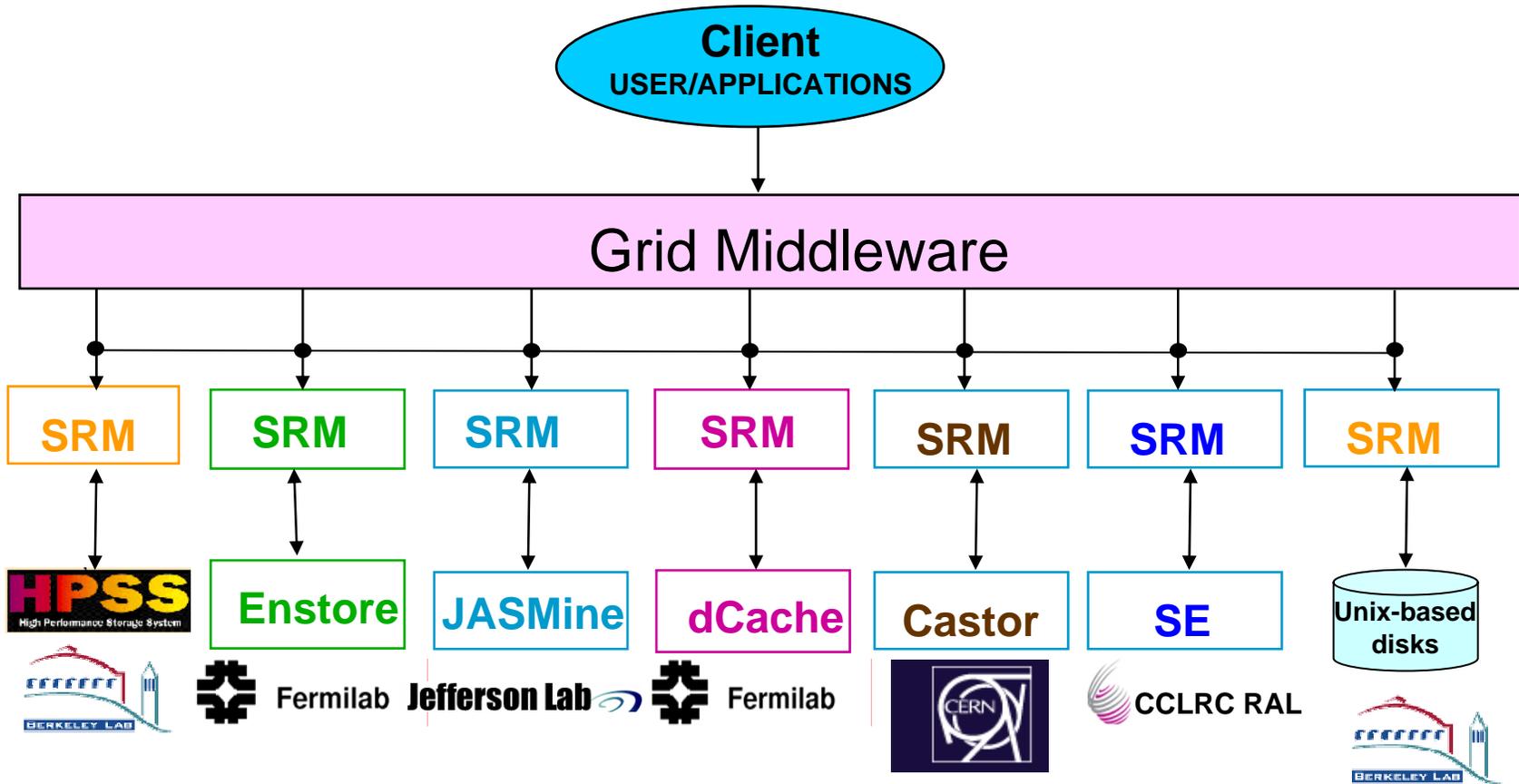
- ❖ Suppose you want to run a job on your local machine
 - Need to allocate space
 - Need to bring all input files
 - Need to ensure correctness of files transferred
 - Need to monitor and recover from errors
 - What if files don't fit space? Need to manage file streaming
 - Need to remove files to make space for more files
- ❖ Now, suppose that the machine and storage space is a shared resource
 - Need to do the above for many users
 - Need to enforce quotas
 - Need to ensure fairness of scheduling users

- ❖ Now, suppose you want to do that on a WAN
 - Need to access a variety of storage systems
 - mostly remote systems, need to have access permission
 - Need to have special software to access mass storage systems
- ❖ Now, suppose you want to run distributed jobs on the WAN
 - Need to allocate remote spaces
 - Need to move (stream) files to remote sites
 - Need to manage file outputs and their movement to destination site(s)

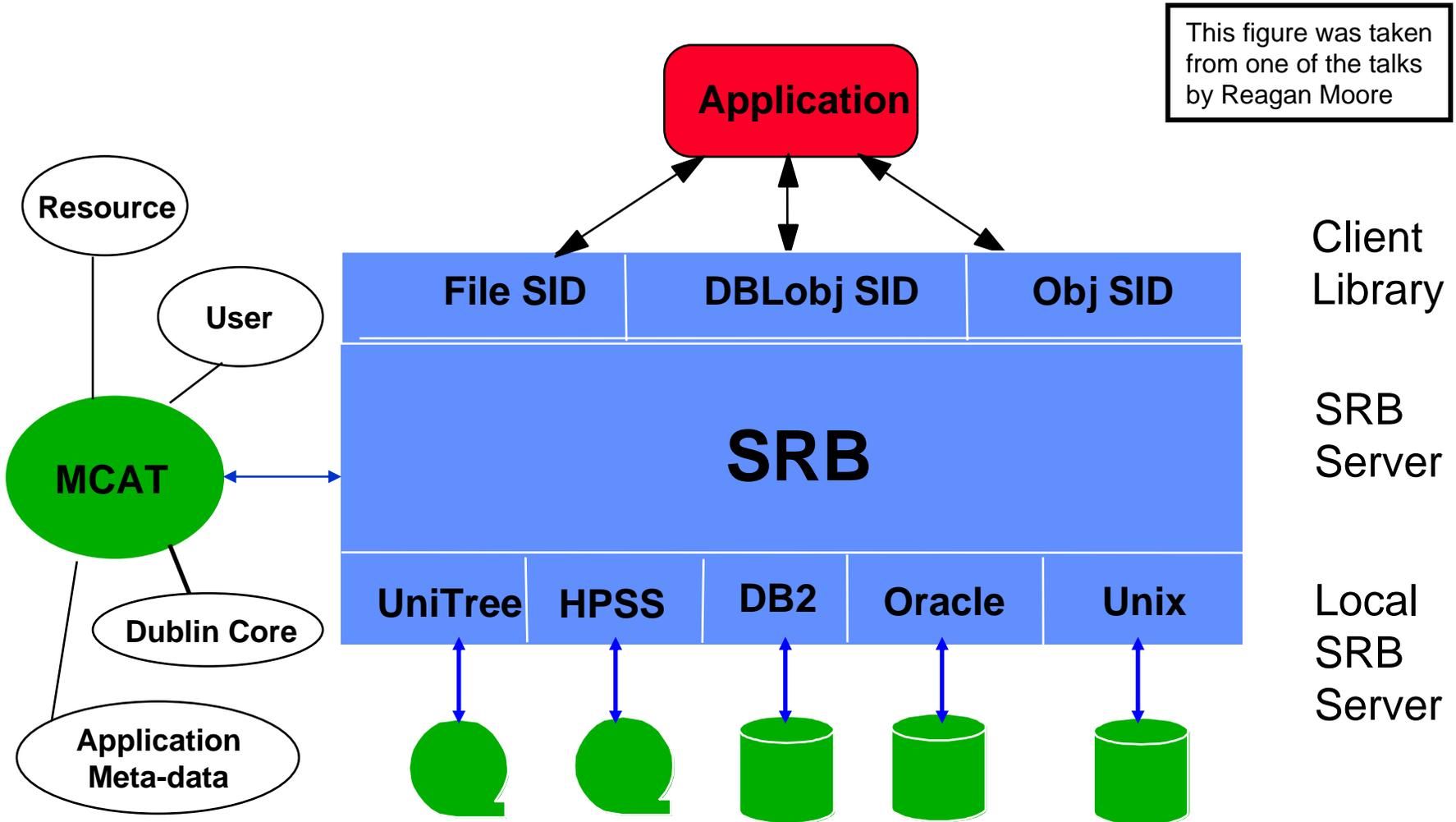
Ubiquitous and Transparent Data Access and Data Sharing



Interoperability of SRMs

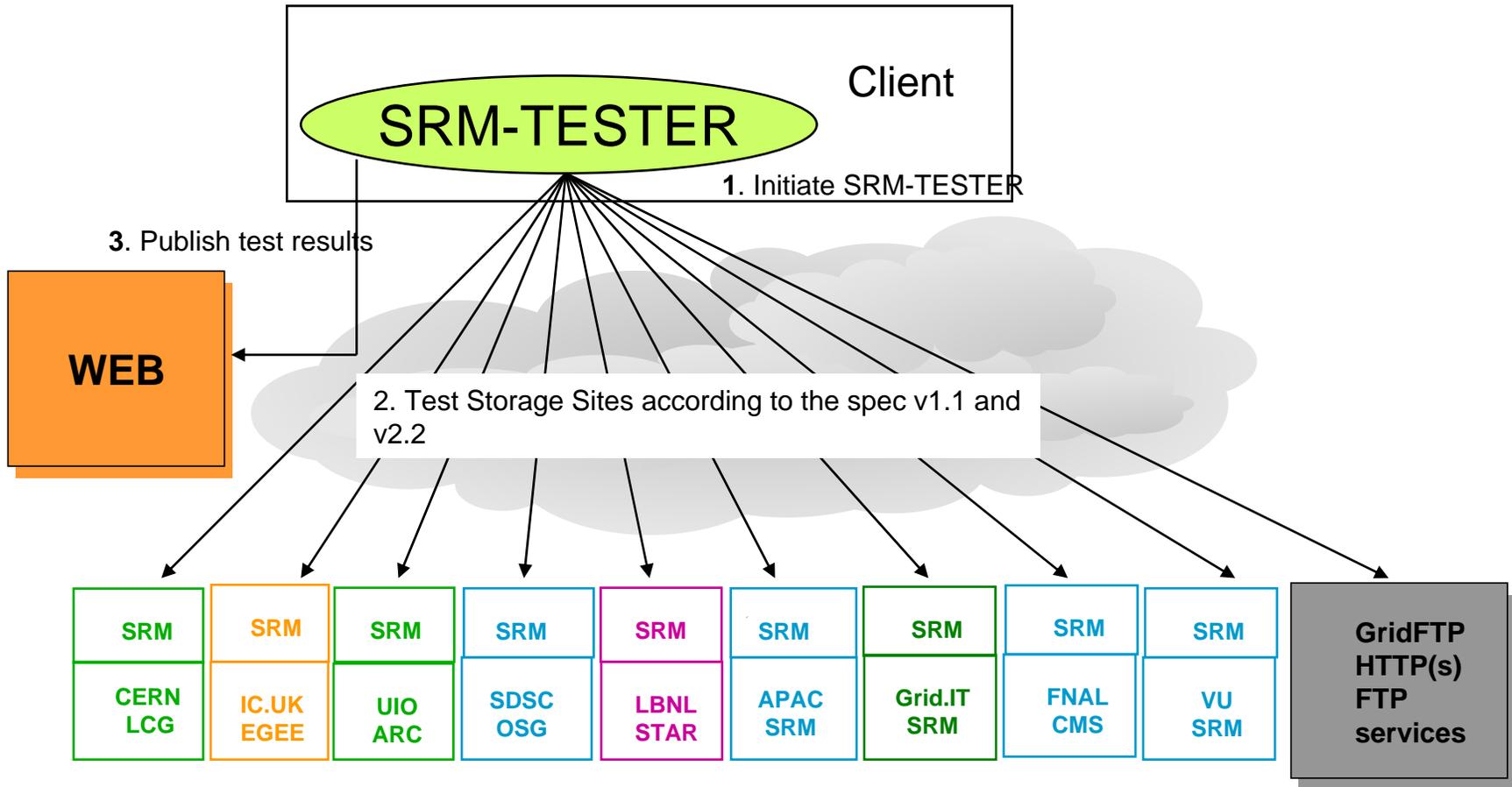


SDSC Storage Resource Broker – Grid Middleware



- ❖ Storage Resource Broker (SRB)
 - Very successful product from SDSC, has long history
 - Is a centralized solution where all requests go to a central server that includes a metadata catalog (MCAT)
 - Developed by a single institution
- ❖ Storage Resource Management (SRM)
 - Based on open standard
 - Developed by multiple institutions for their storage systems
 - Designed for interoperation of heterogeneous storage systems
- ❖ Features of SRM that SRB does not deal with
 - Managing storage space dynamically based client's request
 - Managing content of space based on lifetime controlled by client
 - Support for file streaming by pinning and releasing files
- ❖ Features that SRB that SRM does not deal with
 - Logical organization of file collections
 - Metadata of files and collections (usage attributes, lineage, etc.)
 - Mapping logical file namespace into physical locations of the files.
- ❖ Several projects now ask for an SRM interface to SRB and vice versa
 - In OGF: activity to bridge these technologies

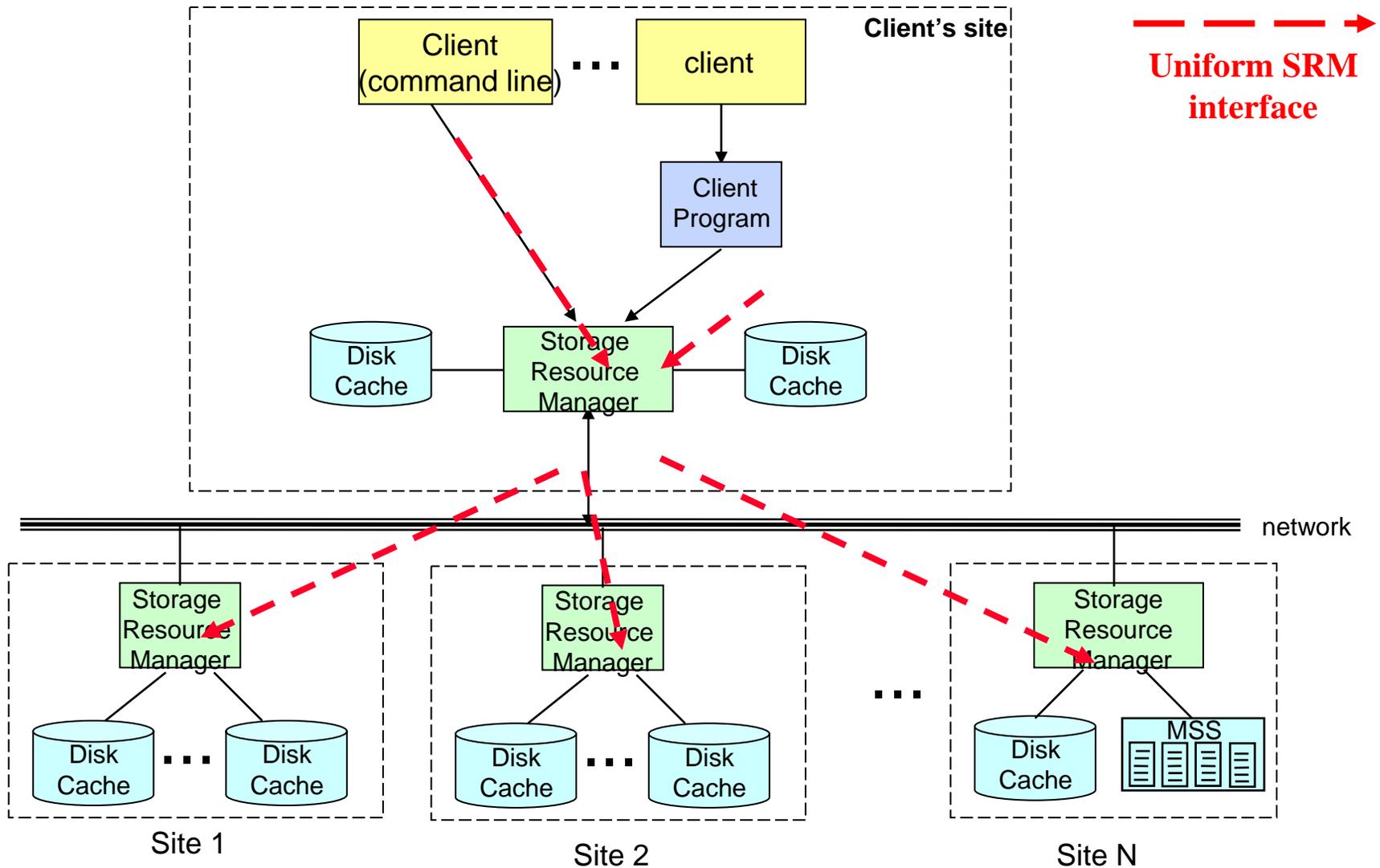
GGF GIN-Data SRM inter-op testing (GGF: Global Grid Forum, GIN: Grid Interoperability Now)



Testing Operations Results

	ping	put	get	Advisory delete	Copy (SRMs)	Copy (gsiftp)
ARC (UIO.NO)	pass	fail	pass	fail	pass	fail
EGEE (IC.UK)	pass	pass	pass	pass	pass	pass
CMS (FNAL.GOV)	pass	pass	pass	pass	pass	pass
LCG/EGEE (CERN)	pass	pass	pass	pass	N.A.	N.A.
OSG (SDSC)	pass	pass	pass	pass	pass	fail
STAR (LBNL)	pass	pass	pass	pass	pass	pass

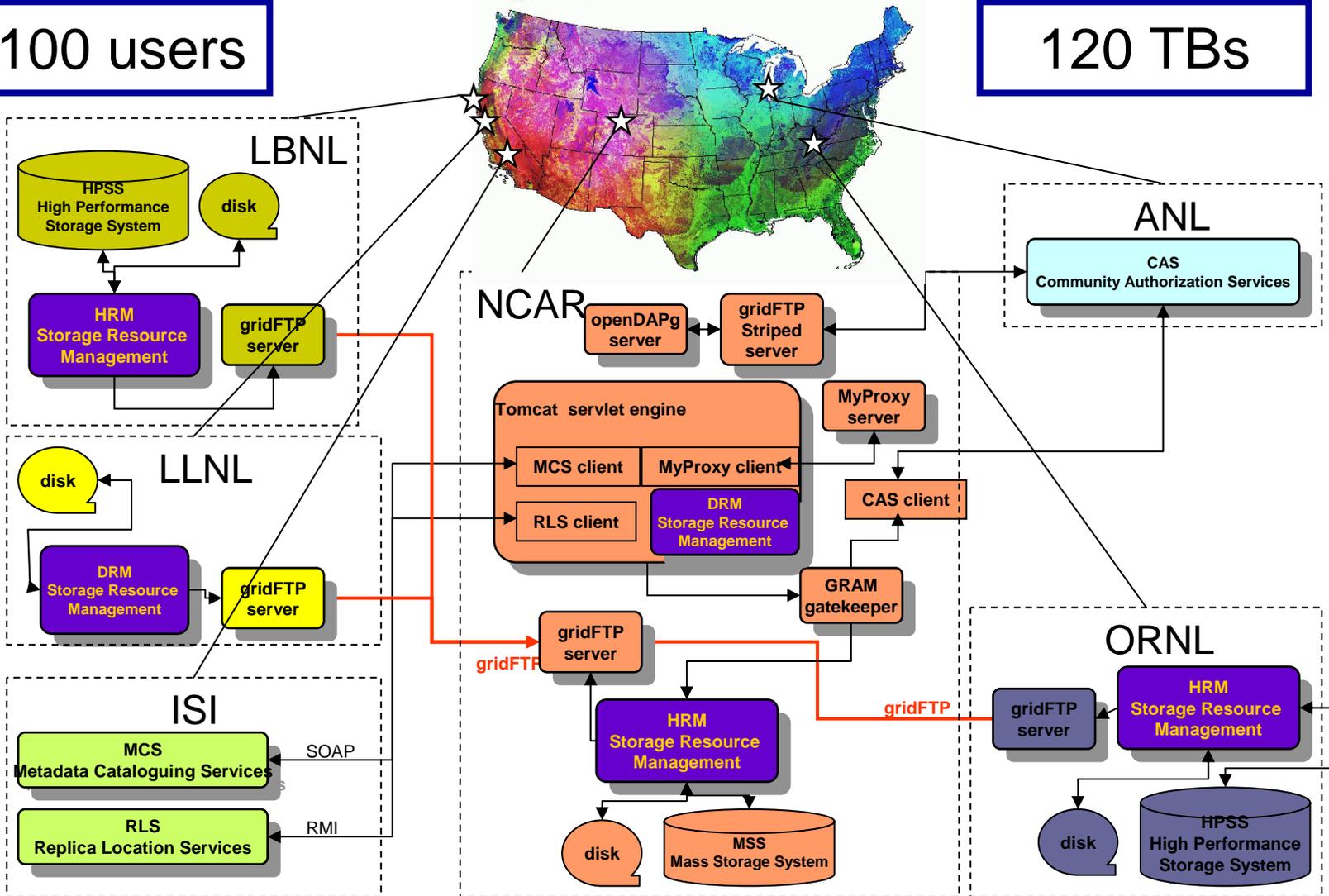
Peer-to-Peer Uniform Interface



Earth Science Grid Analysis Environment

3100 users

120 TBs



History and Partners in SRM Collaboration

- ❖ 5 year of Storage Resource (SRM) Management activity
- ❖ Experience with SRM system implementations
 - Mass Storage Systems:
 - HPSS (LBNL, ORNL, BNL, all using LBNL-SRM),
 - MSS (NCAR - Using LBNL-SRM adaptation),
 - Enstore (Fermi), JasMINE (Jlab), Castor (CERN), Castor (RAL) ...
 - Disk systems:
 - DRM(LBNL), jSRM (Jlab), DPM (CERN), universities ...
 - Combination systems:
 - dCache(Fermi) – sophisticated multi-storage system
 - L-Store (U Vanderbilt) based on Logistical Networking (Using LBNL-SRM adaptation)
 - StoRM – to parallel file systems (ICTP, Trieste, Italy)

Standards for Storage Resource Management

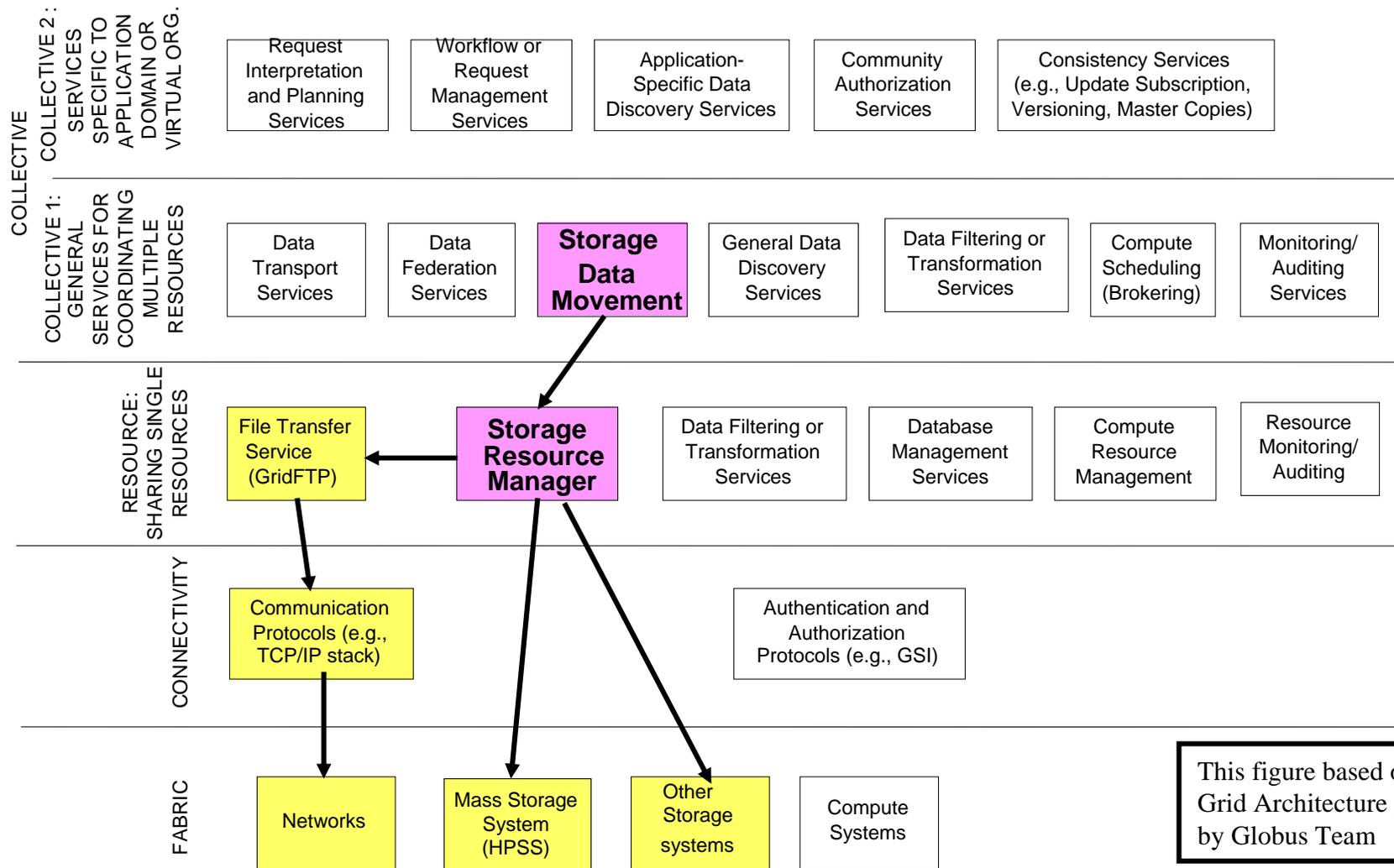
- ❖ Main concepts
 - Allocate spaces
 - Get/put files from/into spaces
 - Pin files for a lifetime
 - Release files and spaces
 - Get files into spaces from remote sites
 - Manage directory structures in spaces
 - SRMs communicate as peer-to-peer
 - Negotiate transfer protocols

- ❖ LBNL leading the **SRM collaboration**
 - a standard common interface was developed and is evolving
 - Institutions involved: LBNL, FNAL, JTNAF, BNL (US), CERN, RAL (Europe)
 - This standard was now adapted by the international WLCG collaboration.
- ❖ SRMs have been used **in production**
 - several facilities including BNL, NERSC, FNAL, CERN, TJNAF, ORNL and NCAR, and other facilities in Europe and Asia.
- ❖ LBNL's SRMs have been used in production over the last few years to support **intensive robust data movement**
 - between BNL to NERSC for the STAR experiment
 - about 10,000 files (about 1 TB) per week in an automated fashion
- ❖ LBNL's SRMs have been used in production in the **Earth Systems Grid (ESG) Project**
 - provide transparent access from multiple remote storage systems at NERSC, NCAR, ORNL, LLNL, and LANL, including HPSS and NCAR-MSS.
- ❖ LBNL's SRMs have also been used for projects in the **Scientific Data Management (SDM) Center**
 - *GridCollector*
 - With MPI-IO to transparently access files on remote storage systems

DataMover

Perform “`rcp -r directory`” on the WAN

SRMs Supports Data Movement Between Storage Systems



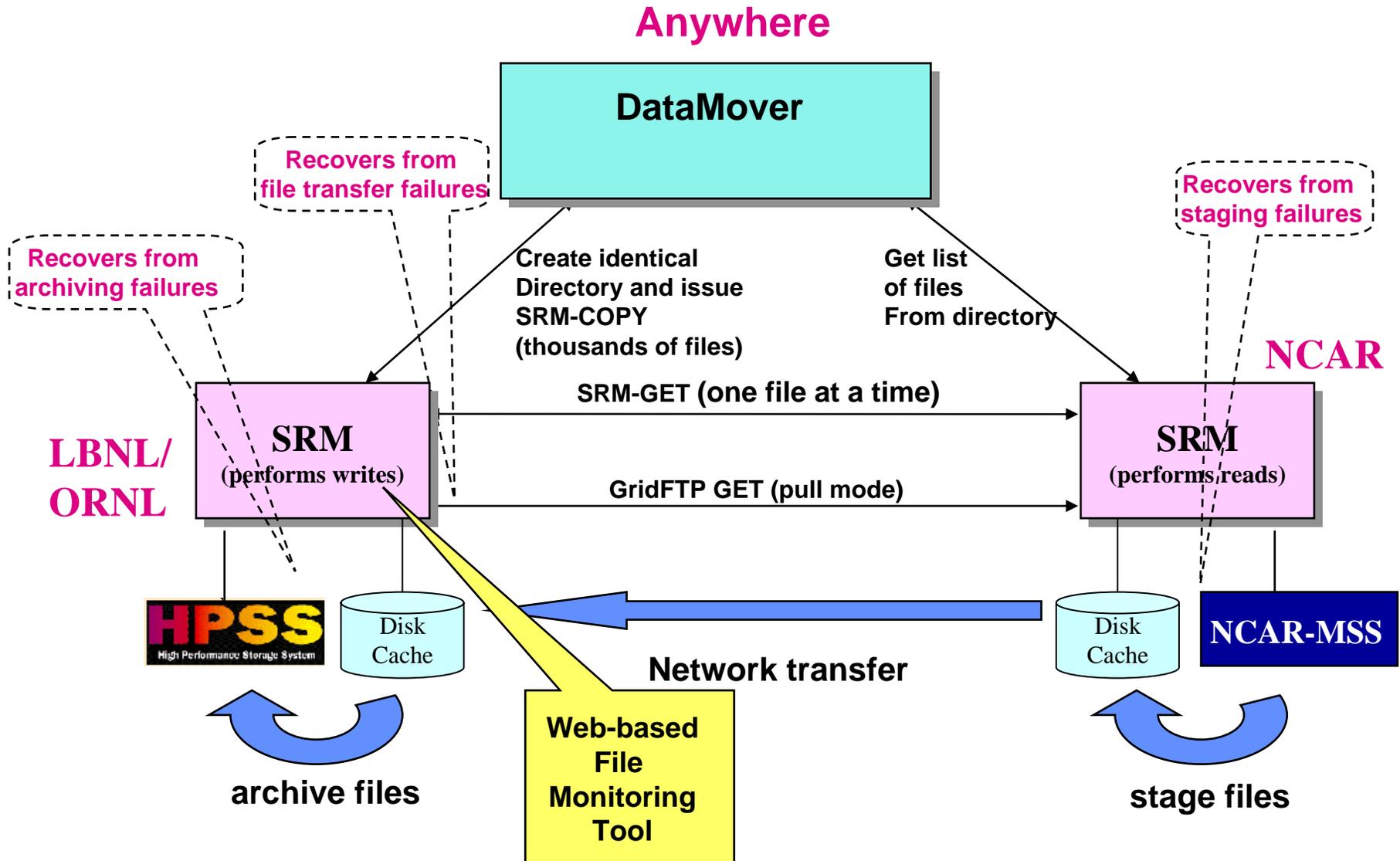
This figure based on the Grid Architecture paper by Globus Team

Massive Robust File Replication



- ❖ Multi-File Replication – why is it a problem?
 - Tedious task – many files, repetitious
 - Lengthy task – long time, can take hours, even days
 - Error prone – need to monitor transfers
 - Error recovery – need to restart file transfers
 - Stage and archive from MSS – limited concurrency, down time, transient failures
 - Commercial MSS – HPSS at NERSC, ORNL, ...,
 - Legacy MSS – MSS at NCAR
 - Independent MSS – Castor (CERN), Enstore (Fermilab), JasMINE (Jlab)

DataMover: SRMs use in ESG for Robust Multi-file replication



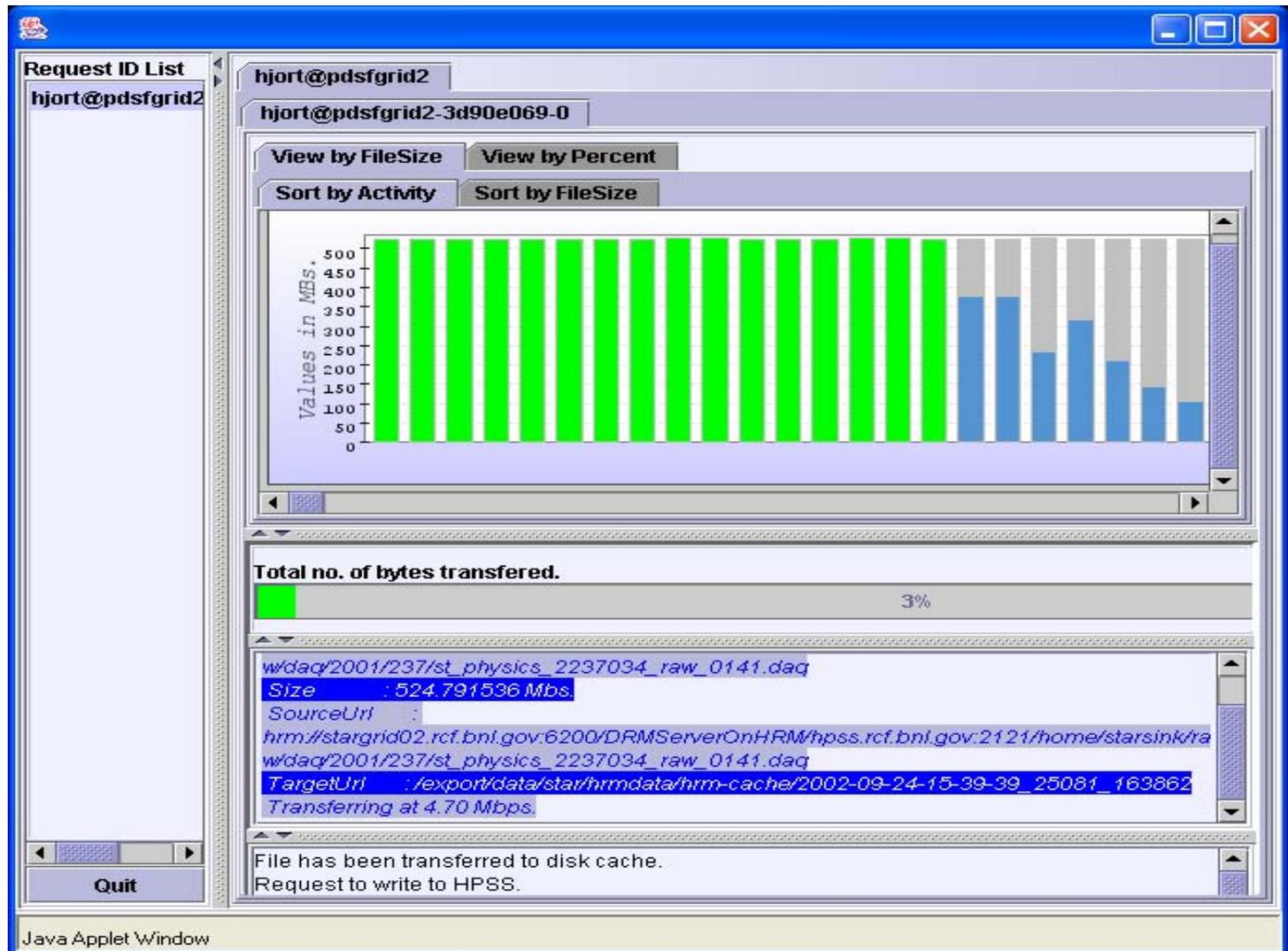
Web-Based File Monitoring Tool

Shows:

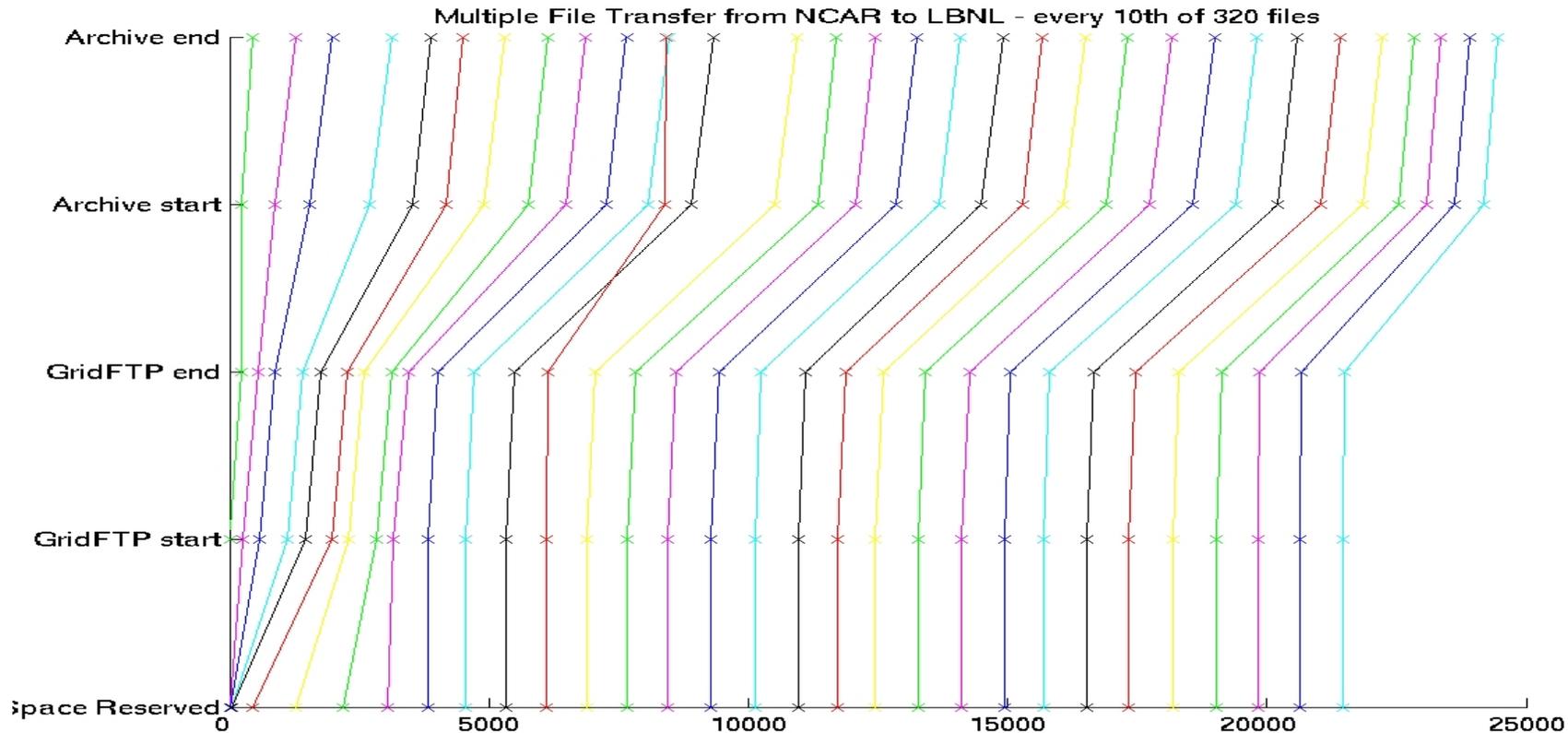
- Files already transferred
- Files during transfer
- Files to be transferred

Also shows for each file:

- Source URL
- Target URL
- Transfer rate

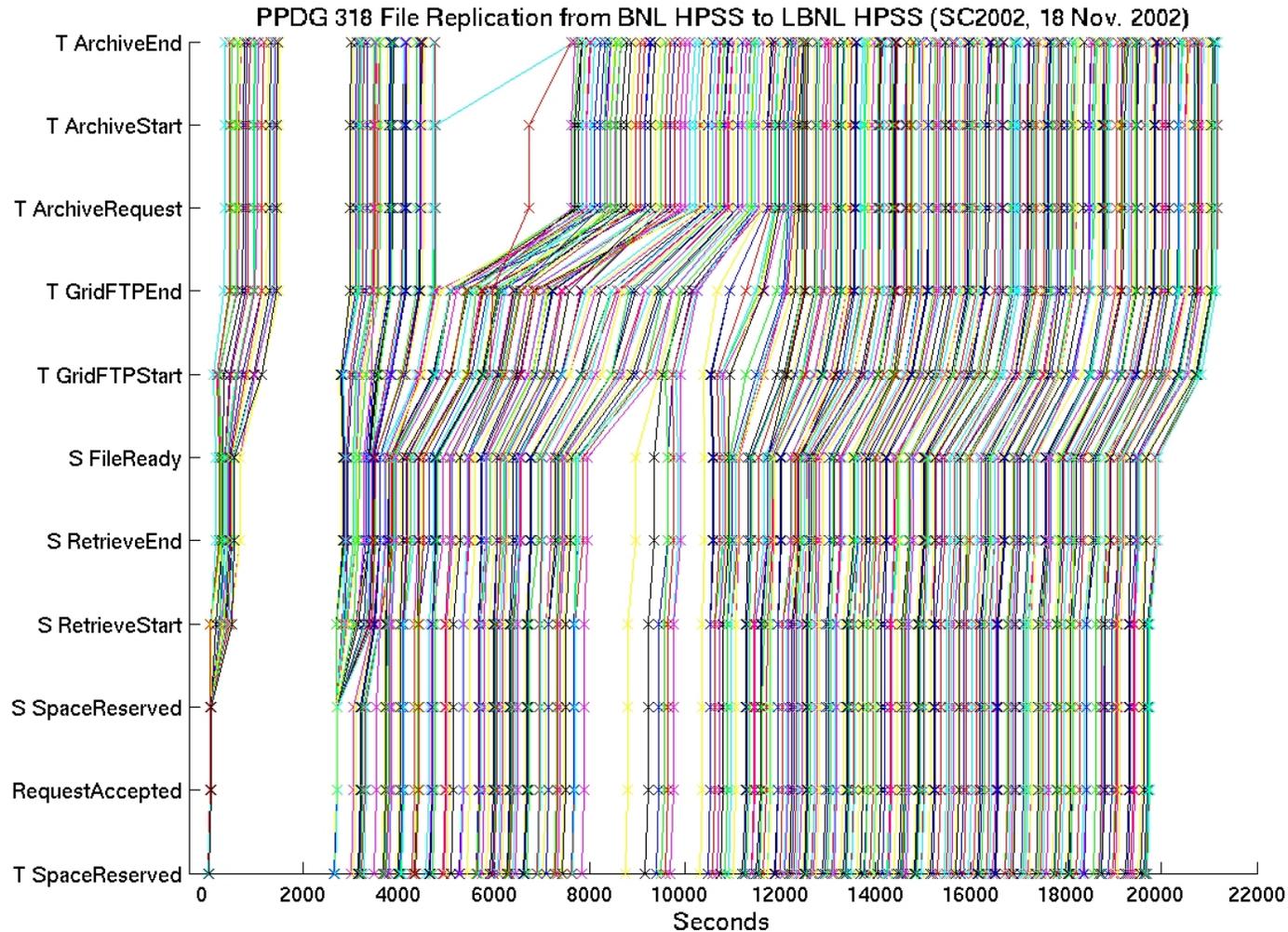


File Tracking Helps to Identify Bottlenecks



Shows that archiving is the bottleneck

File Tracking Shows Recovery From Transient Failures



**Total:
45 GBs**

Robust Multi-File Replication

- ❖ Main results
 - DataMover is being used in production for over three years
 - Moves about 10 TBs a month currently
 - Averages 8 MB/s (64 Mb/sec) over WAN
 - Eliminated person-time to monitor transfer and recover from failures
 - Reduced error rates from about 1% to 0.02% (50 fold reduction)*

* <http://www.ppdg.net/docs/oct04/ppdg-star-oct04.doc>