

# Introduction to Scientific Visualization

Kelly Gaither

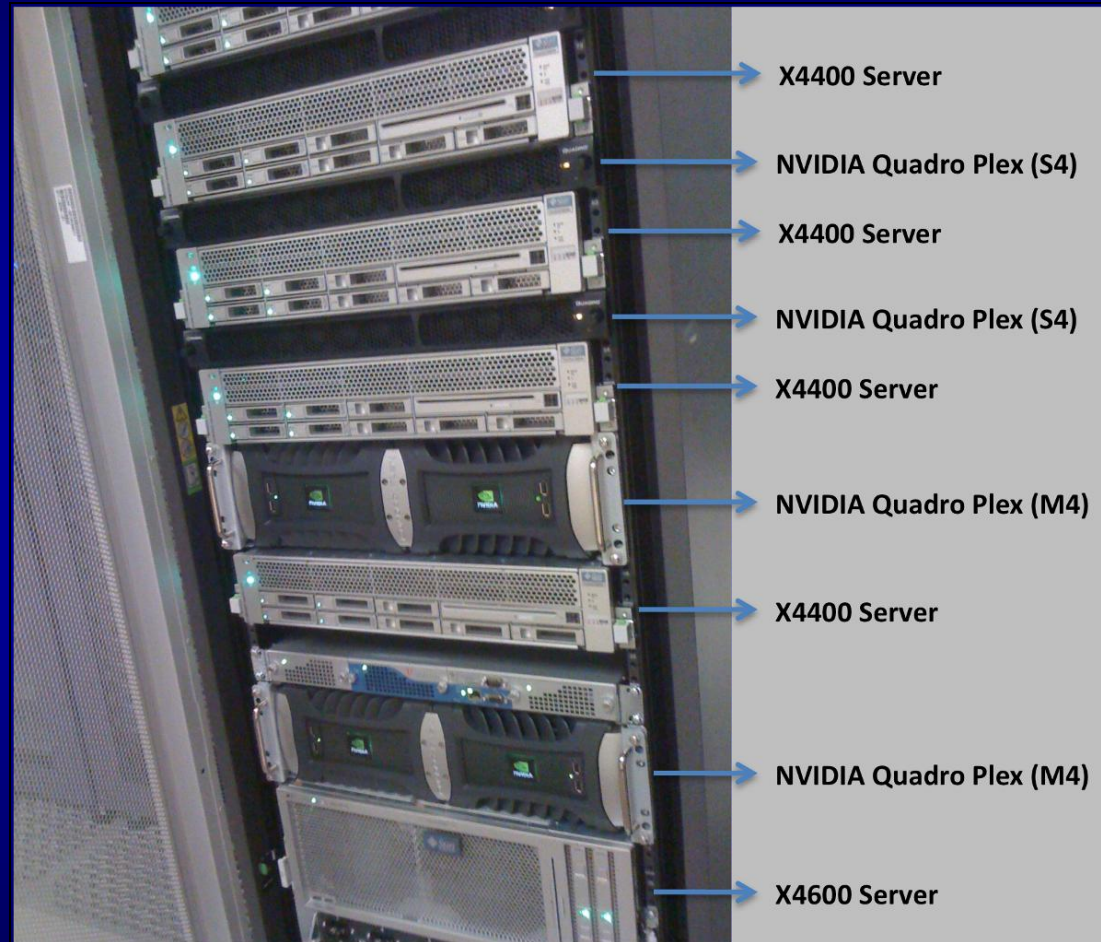
September 2, 2010

# Longhorn Visualization and Data Analysis

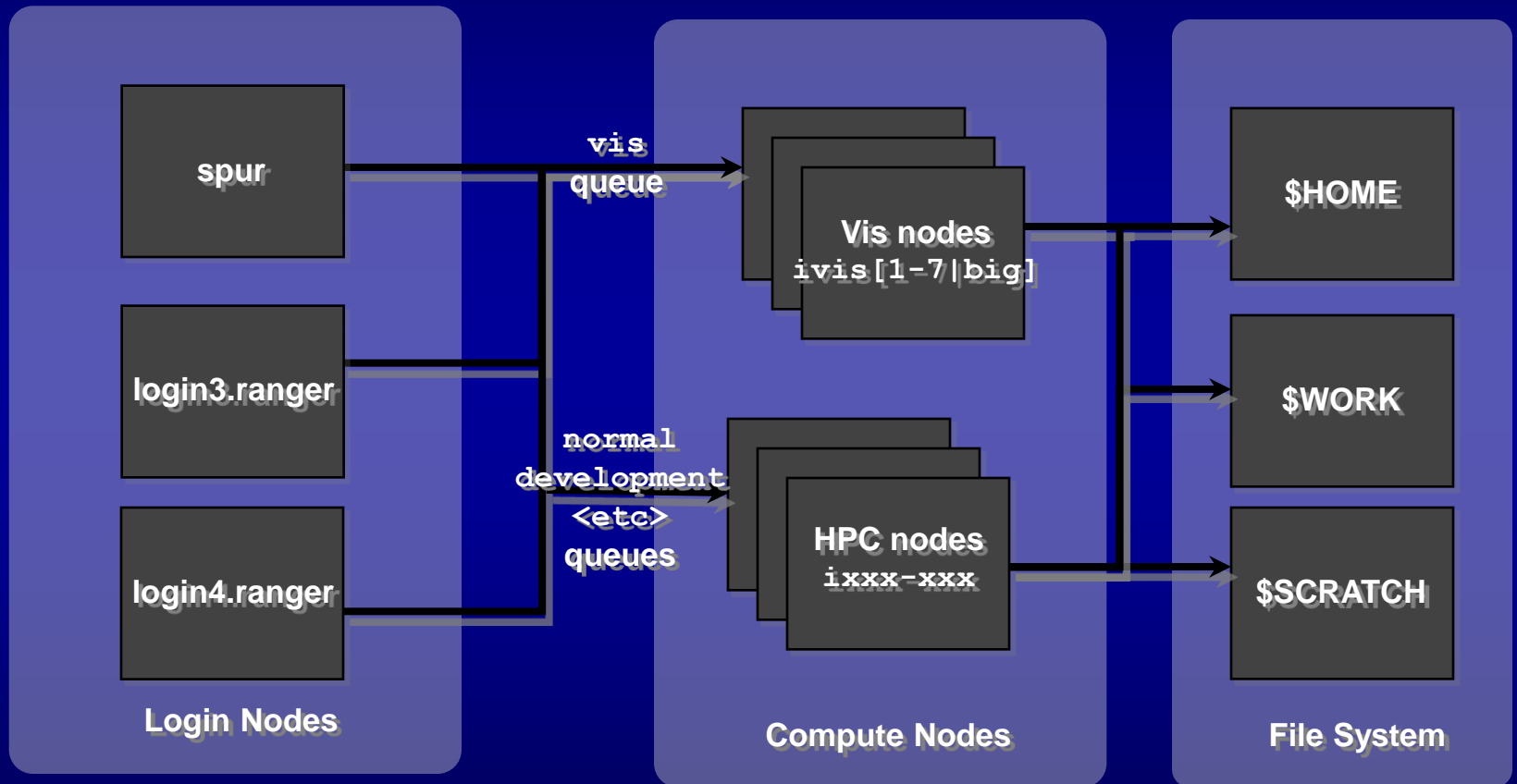
- In November 2008, NSF accepted proposals for the Extreme Digital Resources for Science and Engineering
- The Longhorn project was proposed as a next generation response to TeraGrid's growing visualization and data analysis needs

# Spur - Visualization System

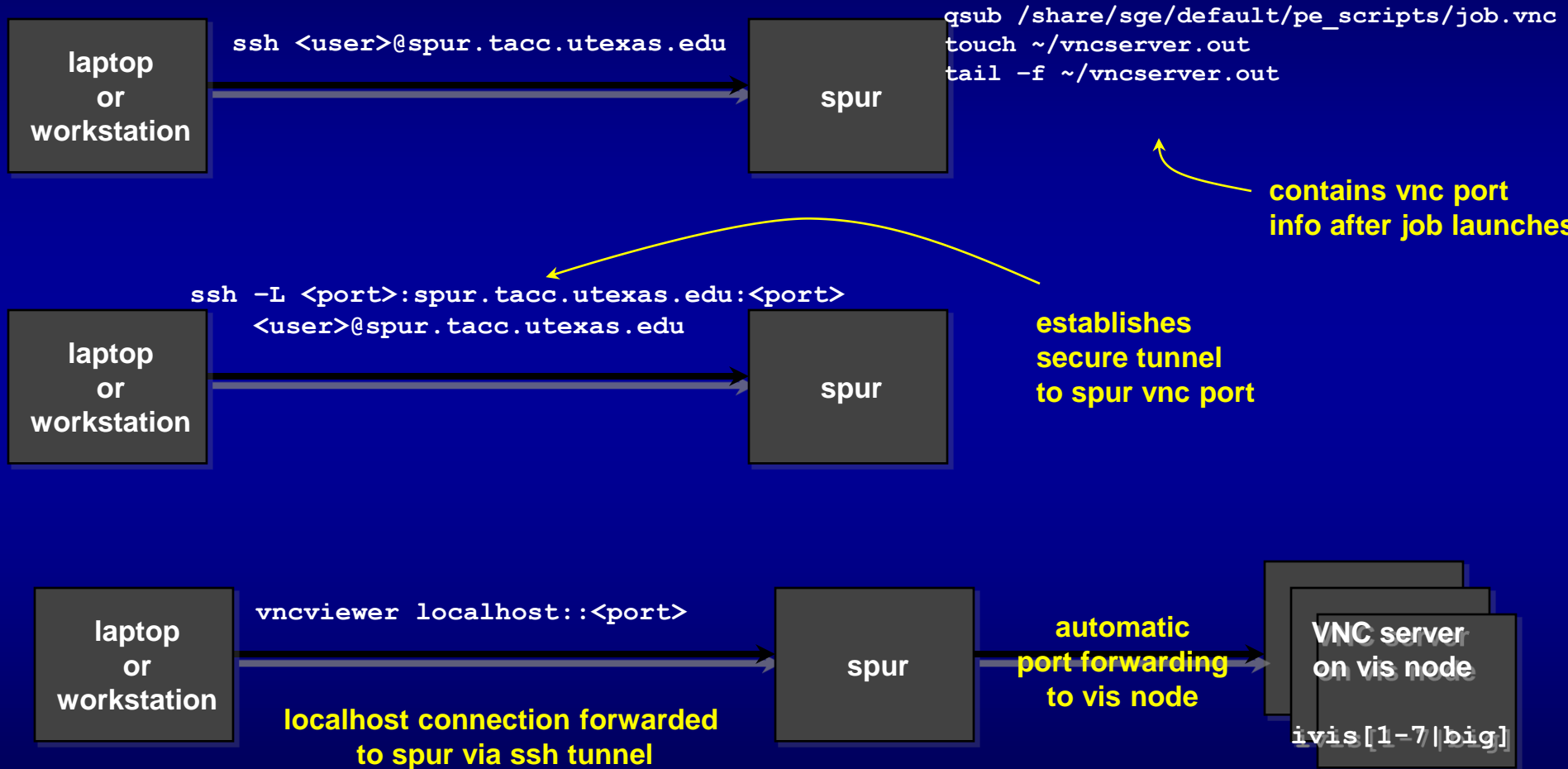
- 128 cores, 1 TB distributed memory, 32 GPUs
- `spur.tacc.utexas.edu` login node, no GPUs  
*don't run apps here!*
- `ivisbig.ranger`  
Sun Fire X4600 server
  - 8 AMD Opteron dual-core CPUs @ 3 GHz
  - 256 GB memory
  - 4 NVIDIA FX5600 GPUs
- `ivis[1-7].ranger`  
Sun Fire X4440 server
  - 4 AMD Opteron quad-core CPUs @ 2.3 GHz
  - 128 GB memory
  - 4 NVIDIA FX5600 GPUs



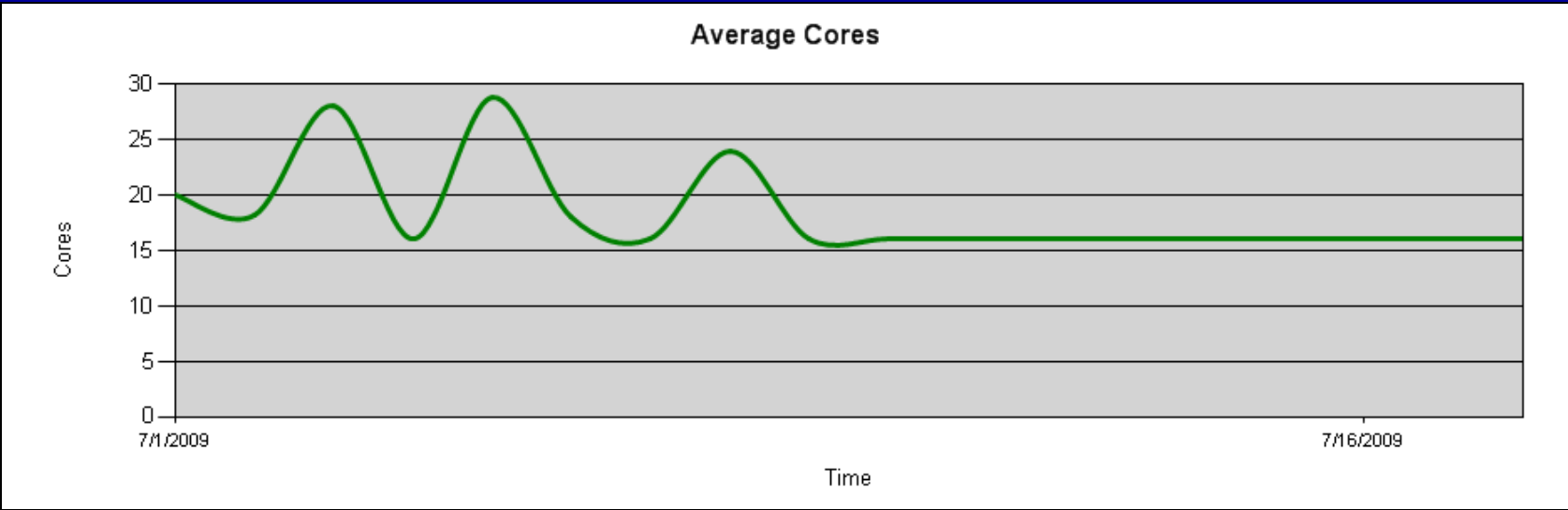
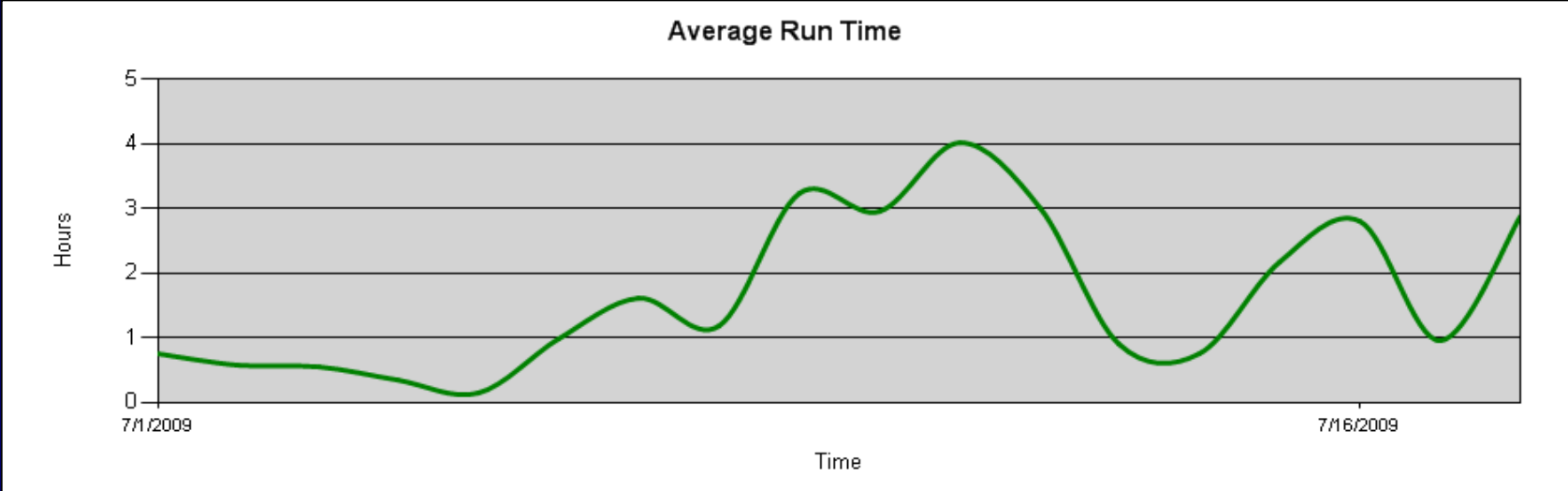
# Spur / Ranger topology



# Connecting to Spur



# Spur Usage

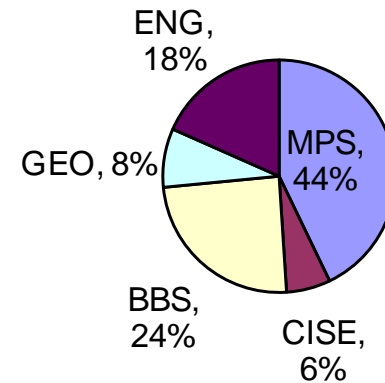


# XD Vis Requirements Analysis

- Surveyed members of the science community via personal interviews and email surveys
- Received ~60 individual responses



NSF Fields of Science Represented



# XD Vis Requirements Analysis

Requirement	% Users Requested
User Support and Consulting	96%
Large-Scale DAV Tools/Resources	39%
Remote/Collaborative DAV Services	27%
Computational Steering	10%
In-simulation DAV Tools	6%
Tools for 3D Measurement and Query	6%
Tools for Multiple Length and Time Scales	6%

(DAV = Data Analysis/Visualization)



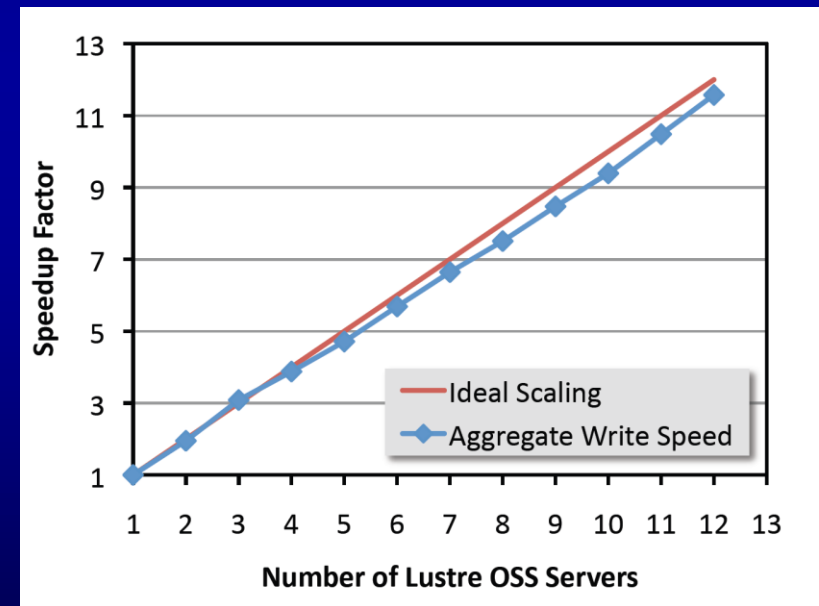
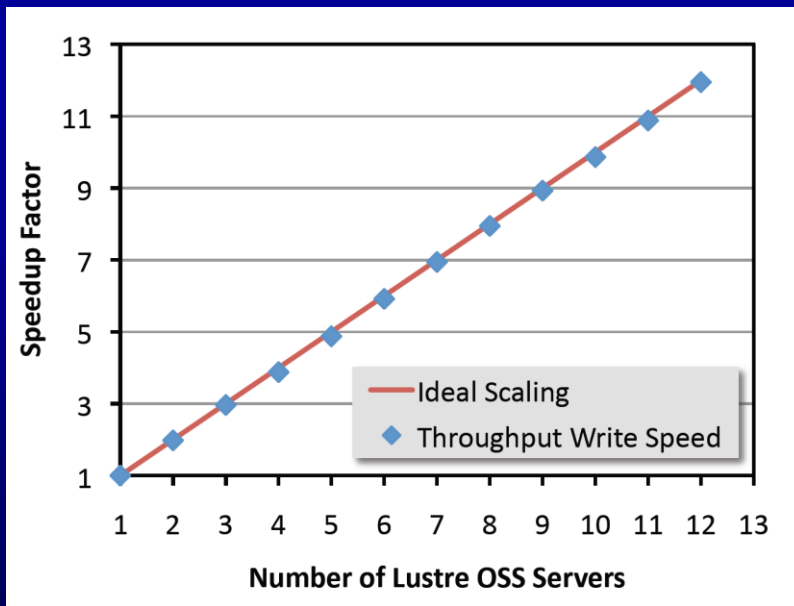
# Longhorn Configuration

**(256 Nodes, 2048 Cores, 512 GPUs, 14.5 TG Aggregate Memory)**

- 256 Dell Quad Core Intel Nehalem Nodes
  - 240 Nodes
    - Dual socket, quad core per socket: 8 cores/node
    - 48 GB shared memory/node (6 GB/core)
    - 73 GB Local Disk
    - 2 Nvidia GPUs/node (FX 5800 - 4GB RAM)
  - 16 Nodes
    - Dual socket, quad core per socket: 8 cores/node
    - 144 GB shared memory/node (18 GB/core)
    - 73 GB Local Disk
    - 2 Nvidia GPUs/node (FX 5800 – 4GB RAM)
  - ~14.5 TB aggregate memory
- QDR InfiniBand Interconnect
- Direct Connection to Ranger's Lustre Parallel File System
- 10G Connection to 210 TB Local Lustre Parallel File System
- Jobs launched through SGE

# Longhorn's Lustre File System (\$SCRATCH)

- OSS's on Longhorn are built on Dell Nehalem Servers Connected to MD10000 Storage Vaults
- 15 Drives Total Configured into 2 Raid5 pairs with a Wandering Spare
- Peak Throughput Speed of the File System is 5.86 GB/sec
- Peak Aggregate Speed of the File System is 5.43 GB/sec



# Longhorn Partners and Roles:

- TACC (Kelly Gaither – PI)
  - Longhorn machine deployment
  - User support
  - Visualization and Data Analysis portal development
  - Software/Tool development
- NCAR (John Clyne – CoPI)
  - User support
  - VAPOR Enhancements
- University of Utah (Valerio Pascucci – CoPI, Chuck Hansen)
  - User support
  - Software Integration of RTRT and topological analysis


# Longhorn Partners and Roles:

- Purdue University (David Ebert – CoPI)
  - User support
  - Integration of visual analytics software
- UC Davis (Hank Childs – Chief Software Integration Architect)
  - Directly facilitate tools being integrated into the VisIt software suite
- SURA (Linda Akli – MSI Outreach/Broadening Participation)

# Longhorn Usage Modalities:

- Remote/Interactive Visualization
  - Highest priority jobs
  - Remote/Interactive capabilities facilitated through VNC
  - Run on 4 hour time limit
- GPGPU jobs
  - Run on a lower priority than the remote/interactive jobs
  - Run on 12 hour time limit
- CPU jobs with higher memory requirements
  - Run on lowest priority when neither remote/interactive nor GPGPU jobs are waiting in the queue
  - Run 12 hour time limit

# Longhorn User Portal

**TACC**  **Longhorn Visualization Portal** TACCkelly [logout](#)  
No resource selected.

Home Allocations Resources Help Admin

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### Select a Resource

Resource:

Project:

Session type:  VNC  EnVision guided visualization

Number of nodes:

*Note: increasing the number of nodes will only increase performance for parallel applications (e.g. ParaView or VisIt).*

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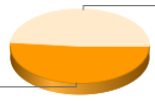
### Available Resources

- Longhorn**

Longhorn (longhorn.tacc.utexas.edu), TACC's Dell XD Visualization Cluster, contains 2048 compute cores, 14.5 TB aggregate memory and 512 GPUs. Longhorn has a QDR InfiniBand interconnect and has an attached Lustre parallel file system. Longhorn is connected by 10GigE to Ranger's Lustre parallel file system thus making it more convenient to work on datasets generated on Ranger. Longhorn has 256 nodes + 2 login nodes, with 240 nodes containing 48GB of RAM, 8 Intel Nehalem cores (@ 2.5 GHz), and 2 NVIDIA Quadro FX 5800 GPUs. Longhorn also has an additional 16 large-memory nodes containing 144GB of RAM, 8 Intel Nehalem cores (@ 2.5 GHz), and 2 NVIDIA Quadro FX 5800 GPUs. For more detailed information on Longhorn, please see the [Longhorn User Guide](#).

**Queue information:**

updated at February 25, 2010, 9:40:11 am ([refresh](#))



The Longhorn queues are open.  
121 nodes available out of 250 total.

```
ACTIVE JOBS-----
JOBID  JOBNAME  USERNAME  STATE  CORE  REMAINING  STARTTIME
-----
6772   ubiq_NVE_5  dlebard   Running  512   11:27:58  Thu Feb 25 09:08:09
6773   vncserverF  pederzan  Running  8     00:00:42  Thu Feb 25 09:10:53
6774   lys_NVE_20  dlebard   Running  512   11:30:58  Thu Feb 25 09:11:09

3 active jobs : 129 of 248 hosts ( 52.02 %)
```

```
WAITING JOBS-----
JOBID  JOBNAME  USERNAME  STATE  CORE  WCLIMIT  QUEUETIME
-----
WAITING JOBS WITH JOB DEPENDENCIES---
```

# Longhorn Queue Structure

## SGE Batch Environment Queues

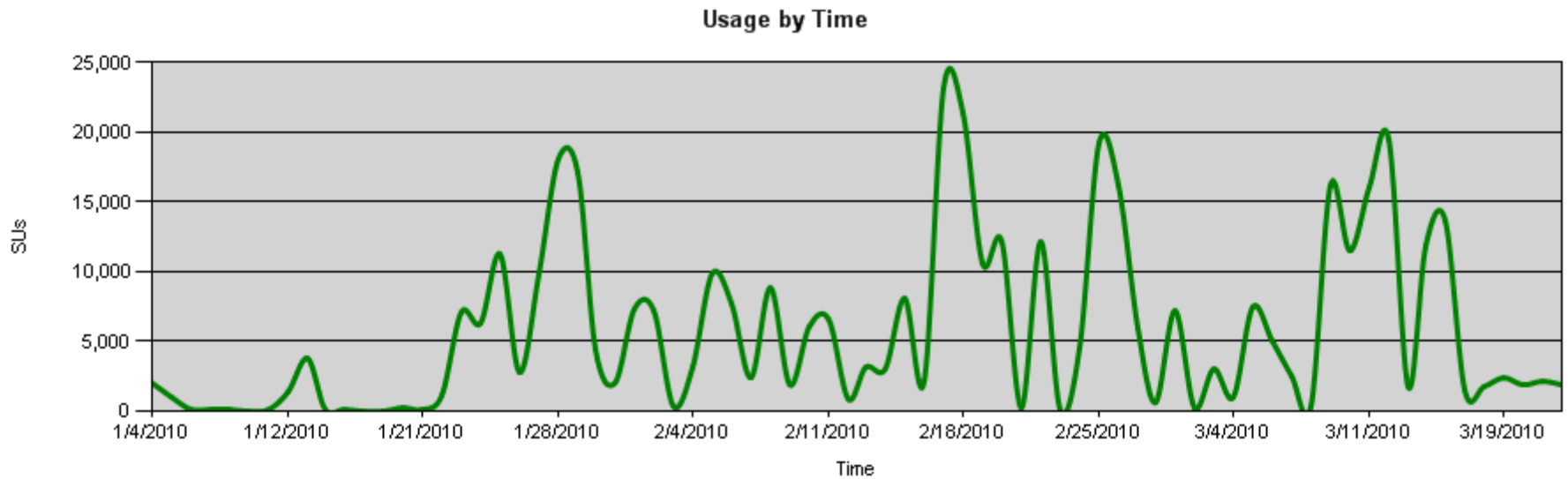
<i>Queue Name</i>	<i>Max Runtime</i>	<i>Max Cores</i>	<i>Node Pool</i>
normal	6 hrs	128	All nodes
long	24 hrs	128	All nodes
largemem	8 hrs	128	16 Large memory nodes
devel	1 hrs	32	8 Nodes
request	---	---	special requests

## Project Types

<i>Type</i>	<i>Purpose</i>	<i>Special Environment Modifications</i>
vis	Visualization jobs	
data	Data Analysis jobs	
gpgpu	GPGPU jobs	disables X server
hpc	HPC jobs	

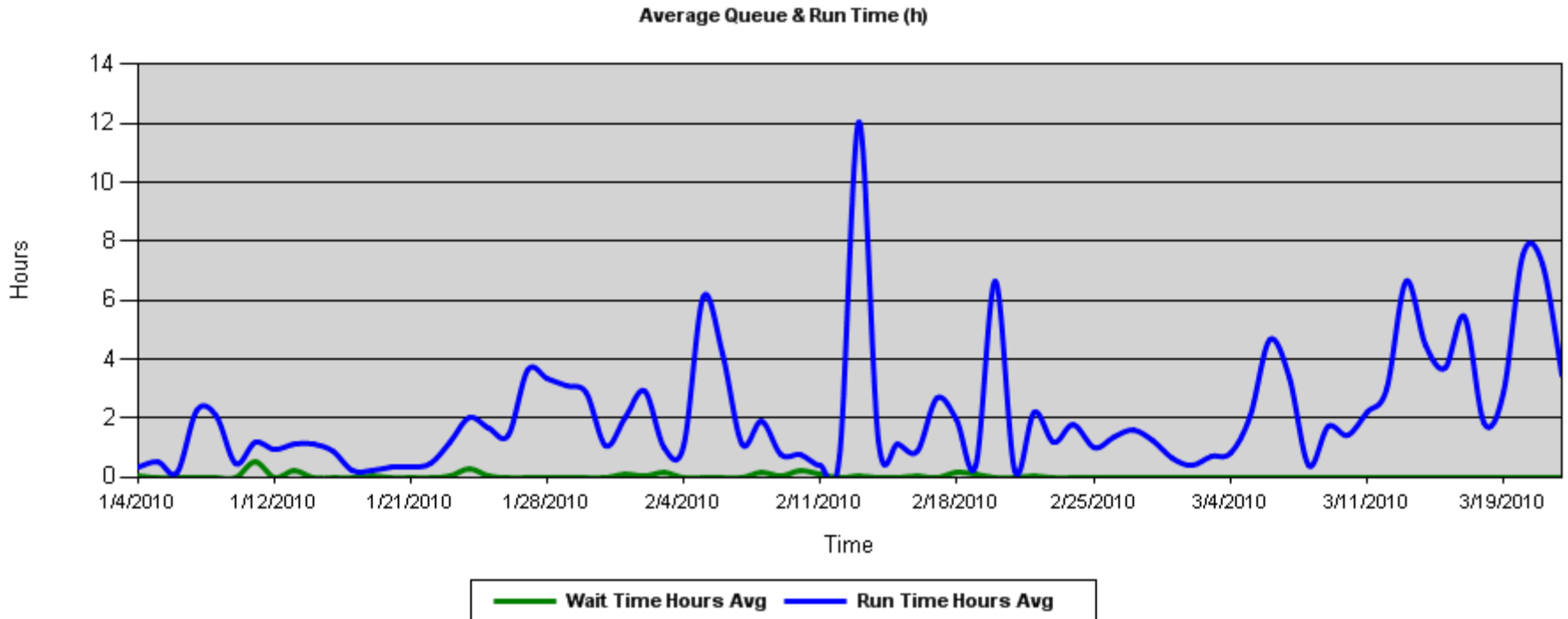
`qsub -q normal -P vis`

# Longhorn Usage

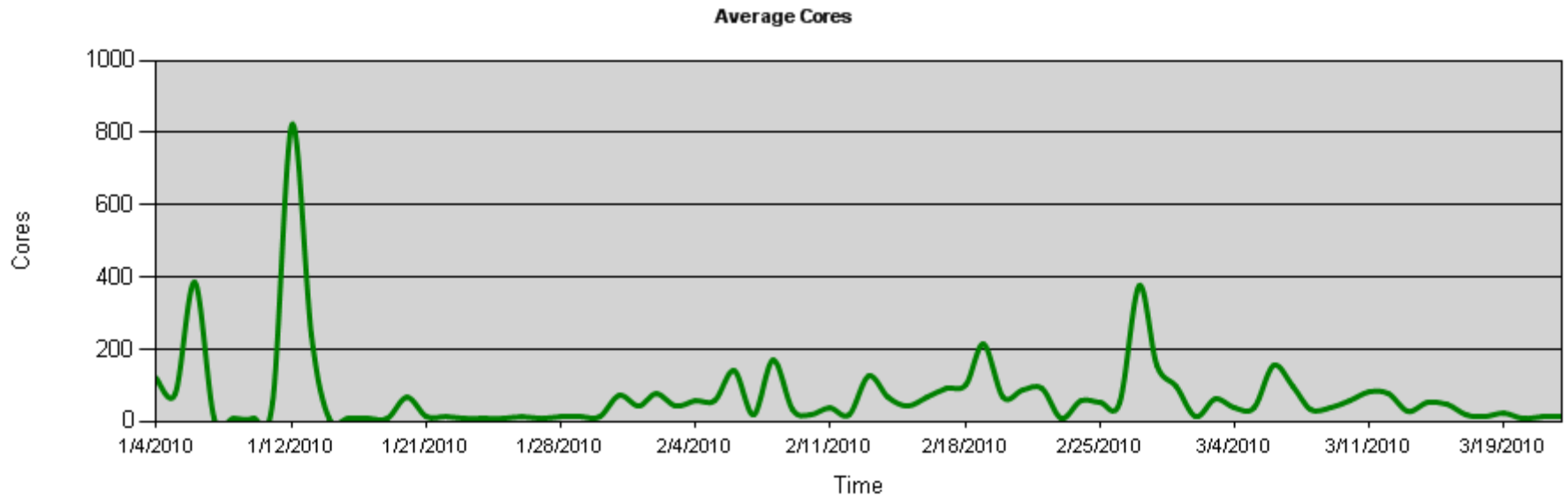




# Longhorn Usage



# Longhorn Usage



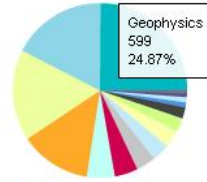
# Longhorn Usage

Field of Science Since Production



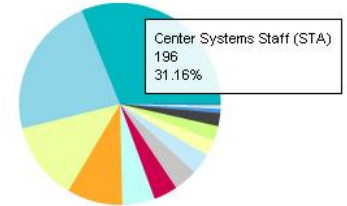
- Advanced Scientific Computing (ASC)
- Research Instrumentation
- Visualization, Graphics and Image Processing
- Unknown
- Special Projects
- Geophysics
- Center Systems Staff (STA)
- Computational Mathematics
- Physical Chemistry
- Condensed Matter Physics
- COMPUTER AND INFORMATION SCIENCE AND ENGINEERING (CISE)
- Fluid, Particulate, and Hyrdraulic Systems
- Training (TRA)
- Tectonics
- MATHEMATICAL AND PHYSICAL SCIENCES (MPS)
- Extragalactic Astronomy and Cosmology
- Software Development
- Seismology
- Cross-Disciplinary Activities (CDA)

Field of Science Last 30 Days



- Geophysics 599 24.87%
- Research Instrumentation
- Special Projects
- Center Systems Staff (STA)
- Condensed Matter Physics
- Visualization, Graphics and Image Processing
- Advanced Scientific Computing (ASC)
- Physical Chemistry
- Fluid, Particulate, and Hyrdraulic Systems
- Tectonics
- COMPUTER AND INFORMATION SCIENCE AND ENGINEERING (CISE)
- Software Development
- MATHEMATICAL AND PHYSICAL SCIENCES (MPS)
- Seismology
- Cross-Disciplinary Activities (CDA)
- Extragalactic Astronomy and Cosmology

Field of Science Last 7 Days



- Center Systems Staff (STA) 196 31.16%
- Special Projects
- Advanced Scientific Computing (ASC)
- Research Instrumentation
- Tectonics
- Software Development
- Physical Chemistry
- Fluid, Particulate, and Hyrdraulic Systems
- Seismology
- Visualization, Graphics and Image Processing
- COMPUTER AND INFORMATION SCIENCE AND ENGINEERING (CISE)
- MATHEMATICAL AND PHYSICAL SCIENCES (MPS)
- Cross-Disciplinary Activities (CDA)
- Extragalactic Astronomy and Cosmology

# Sampling of Current Projects

- Computational Study of Earth and Planetary Materials
- Simulation of Quantum Systems
- Visualization and Analysis of Turbulent Flow
- A probabilistic Molecular Dynamics Optimized for the GPU
- Visualization of Nano-Microscopy
- MURI on Biologically-Inspired Autonomous Sea Vehicles: Towards a Mission Configurable Stealth Underwater Batoid
- Adaptive Multiscale Simulations

# Scientific Visualization

**“The purpose of computing is insight not numbers.”**

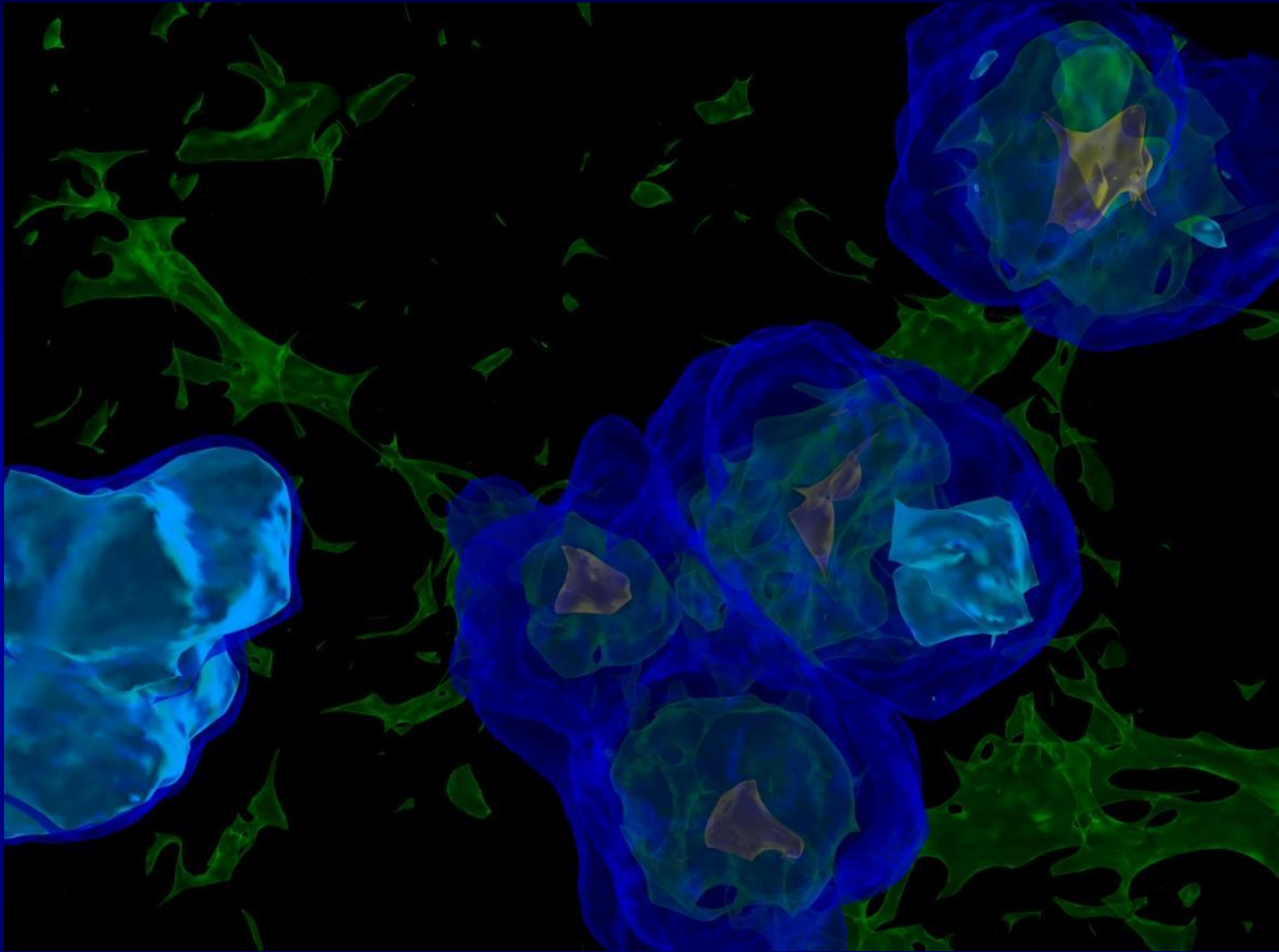
**-- R. W. Hamming (1961)**



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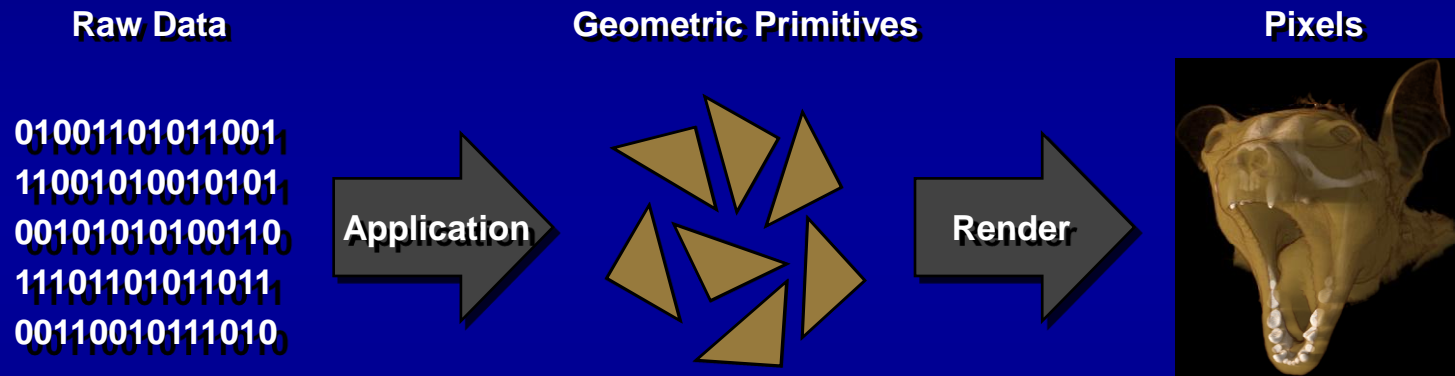




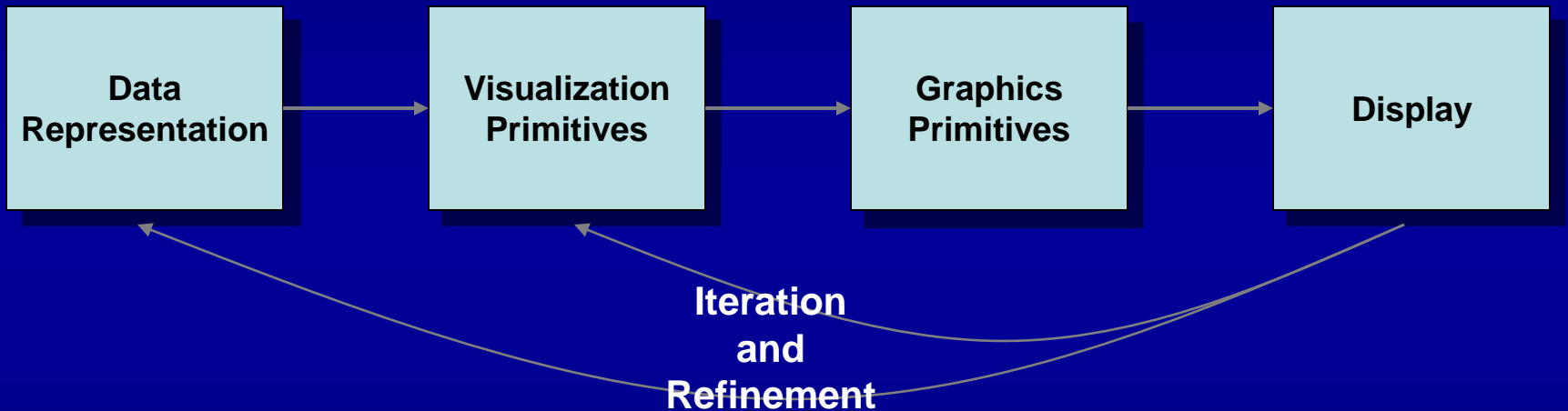




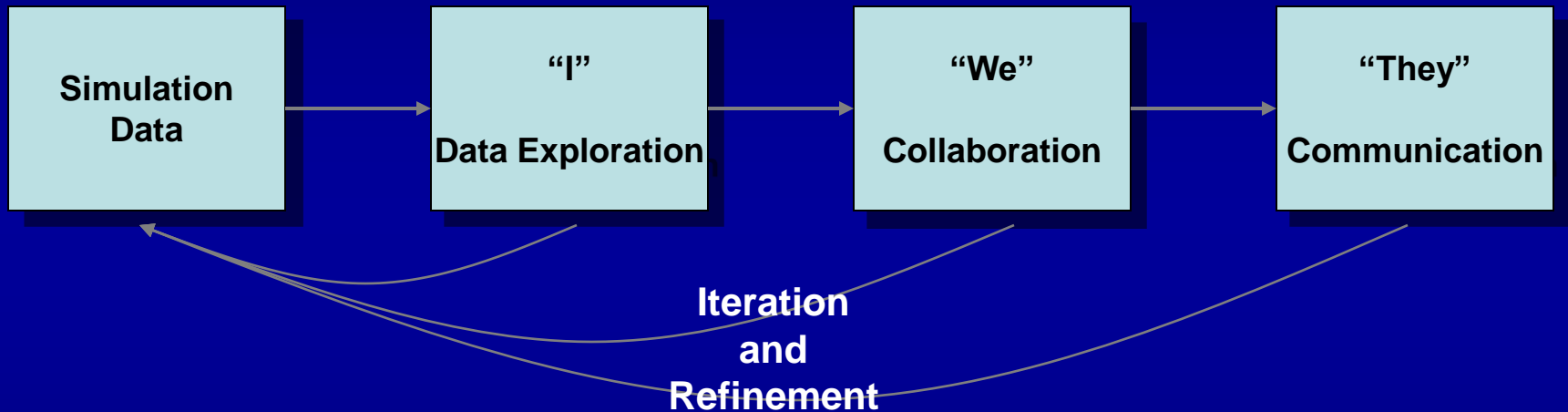
# Visualization Allows Us to “See” the Science



# Getting from Data to Insight



# “I, We, They” Development Path



# Visualization Process Summary

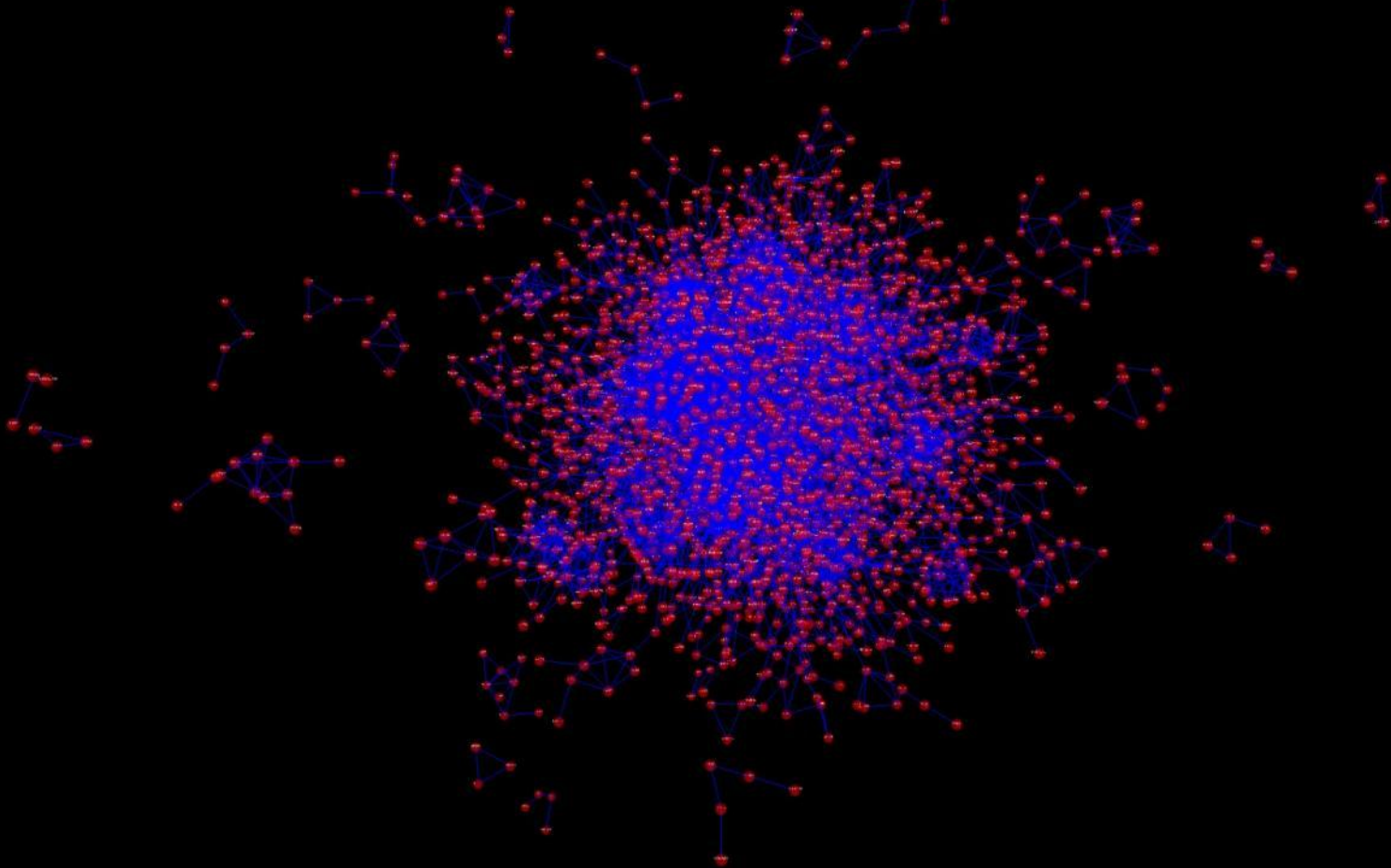
- The primary goal of visualization is *insight*
- A picture is worth not just 1000 words, but potentially tera- or peta-bytes of data
- Larger datasets demand not just visualization, but advanced visualization resources and techniques
- Visualization system technology improves with advances in GPUs and LCD technology
- Visualization software slower to adapt

# Types of Input Data

- Point / Particle
  - N-body simulation
- Regular grid
  - Medical scan
- Curvilinear grid
  - Engineering model
- Unstructured grid
  - Extracted surface

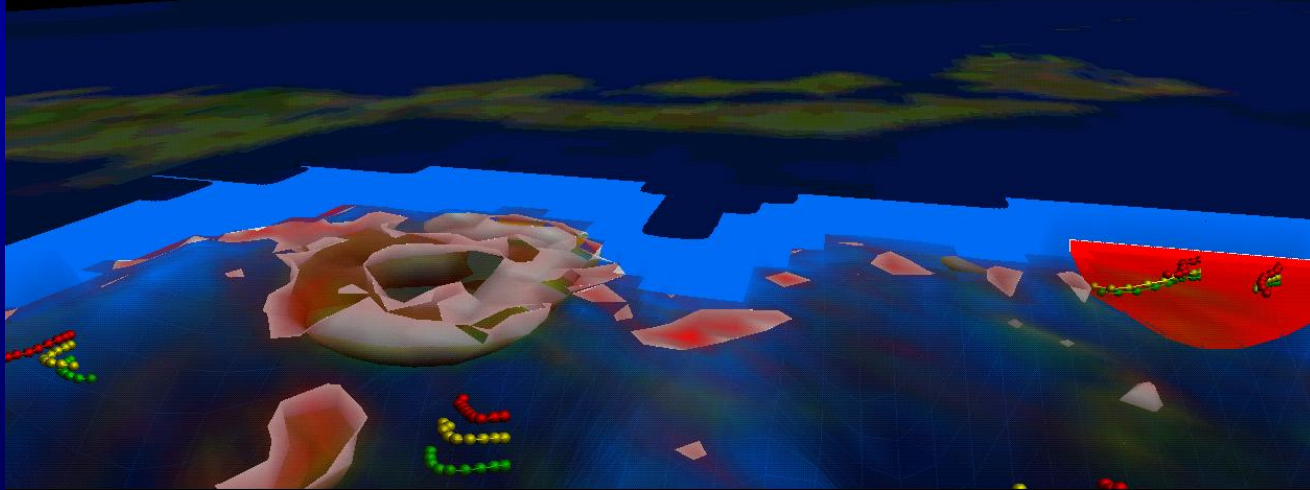
# Types of Input Data

Point – scattered values with no defined structure



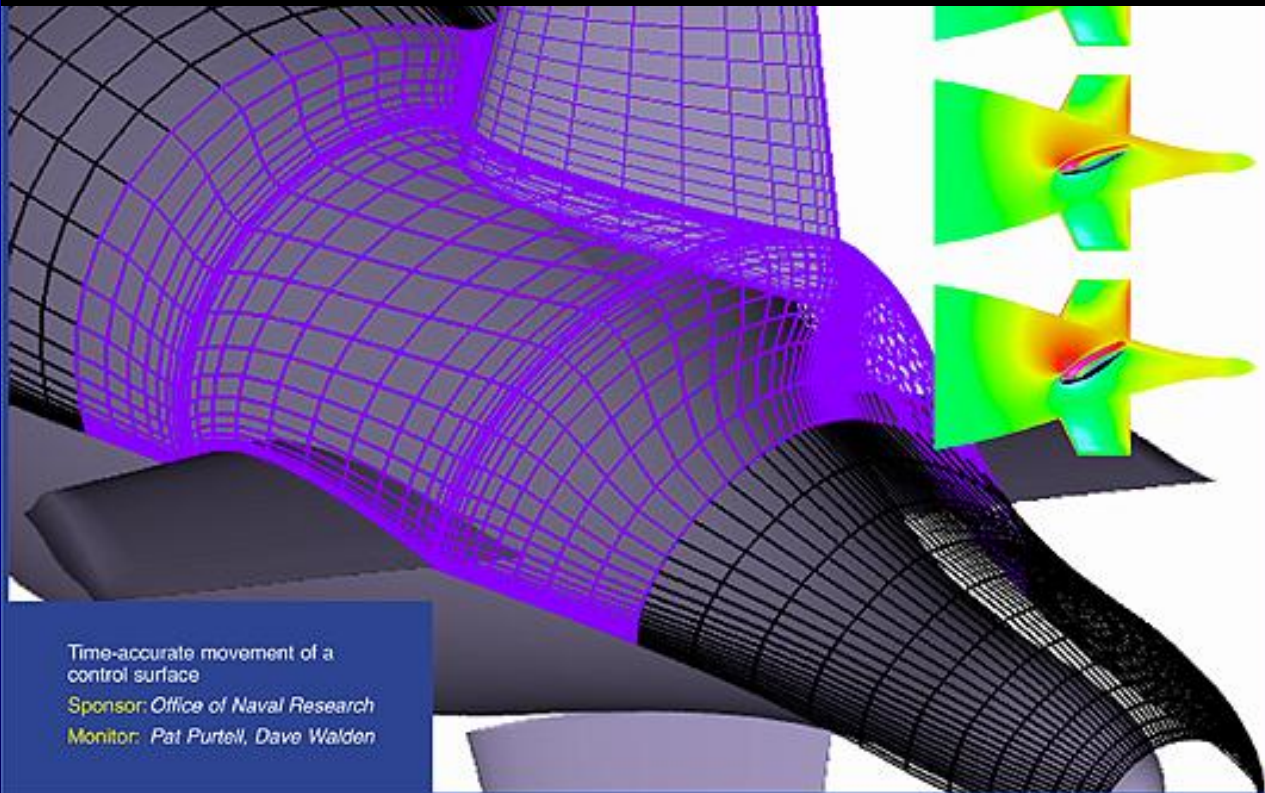
# Types of Input Data

Grid – regular structure, all voxels (cells) are the same size and shape



# Types of Input Data

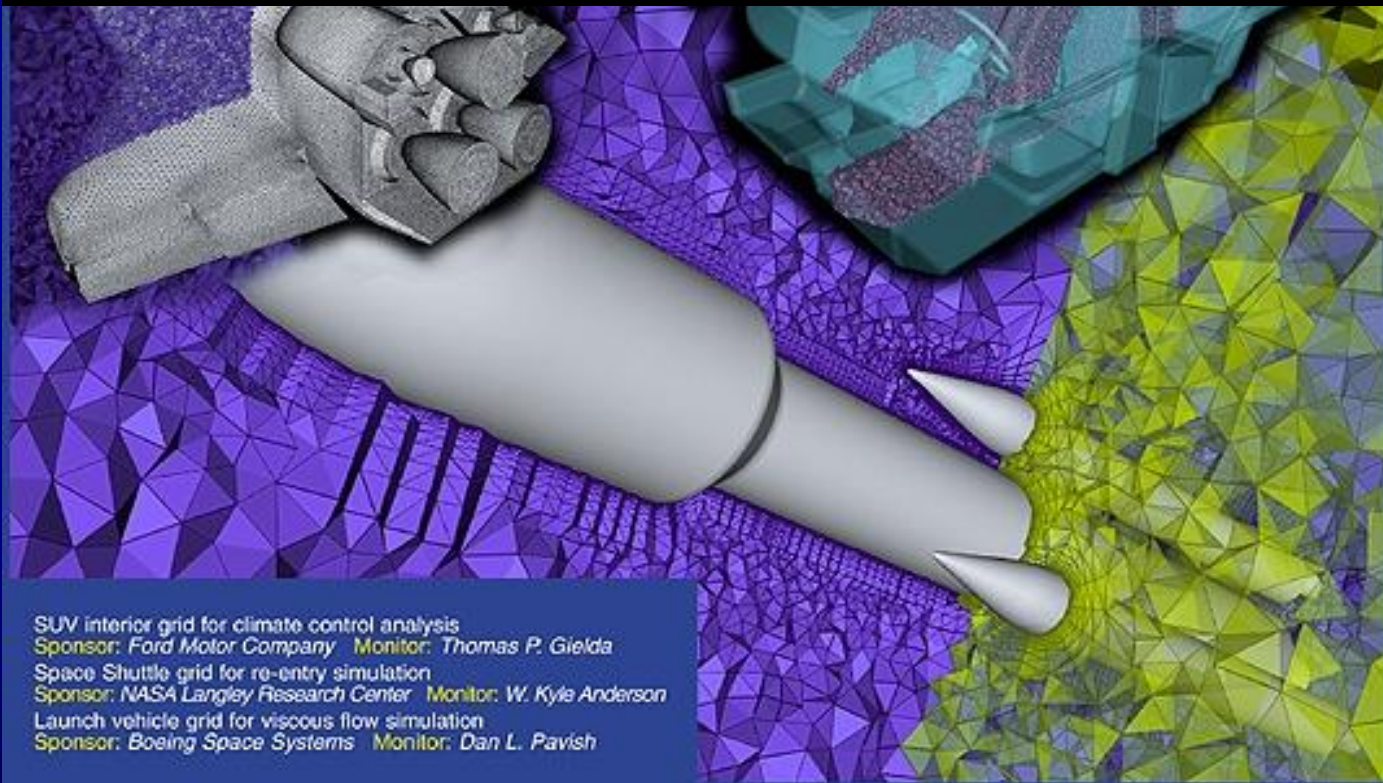
Curvilinear – regularly grided mesh  
shaping function applied





# Types of Input Data

Unstructured grid – irregular mesh typically composed of tetrahedra, prisms, pyramids, or hexahedra.



SUV interior grid for climate control analysis  
Sponsor: Ford Motor Company Monitor: Thomas P. Giolda  
Space Shuttle grid for re-entry simulation  
Sponsor: NASA Langley Research Center Monitor: W. Kyle Anderson  
Launch vehicle grid for viscous flow simulation  
Sponsor: Boeing Space Systems Monitor: Dan L. Pavish

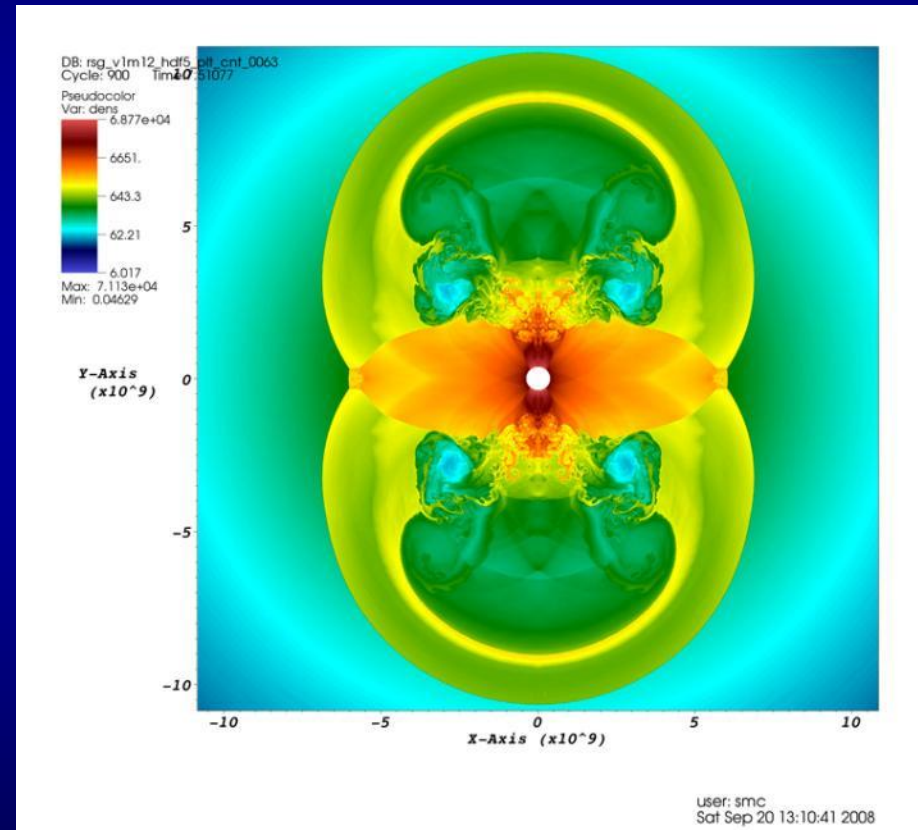
# Visualization Operations

- Surface Shading (Pseudocolor)
- Isosurfacing (Contours)
- Volume Rendering
- Clipping Planes
- Streamlines

# Surface Shading (Pseudocolor)

Given a scalar value at a point on the surface and a color map, find the corresponding color (and opacity) and apply it to the surface point.

Most common operation, often combined with other ops

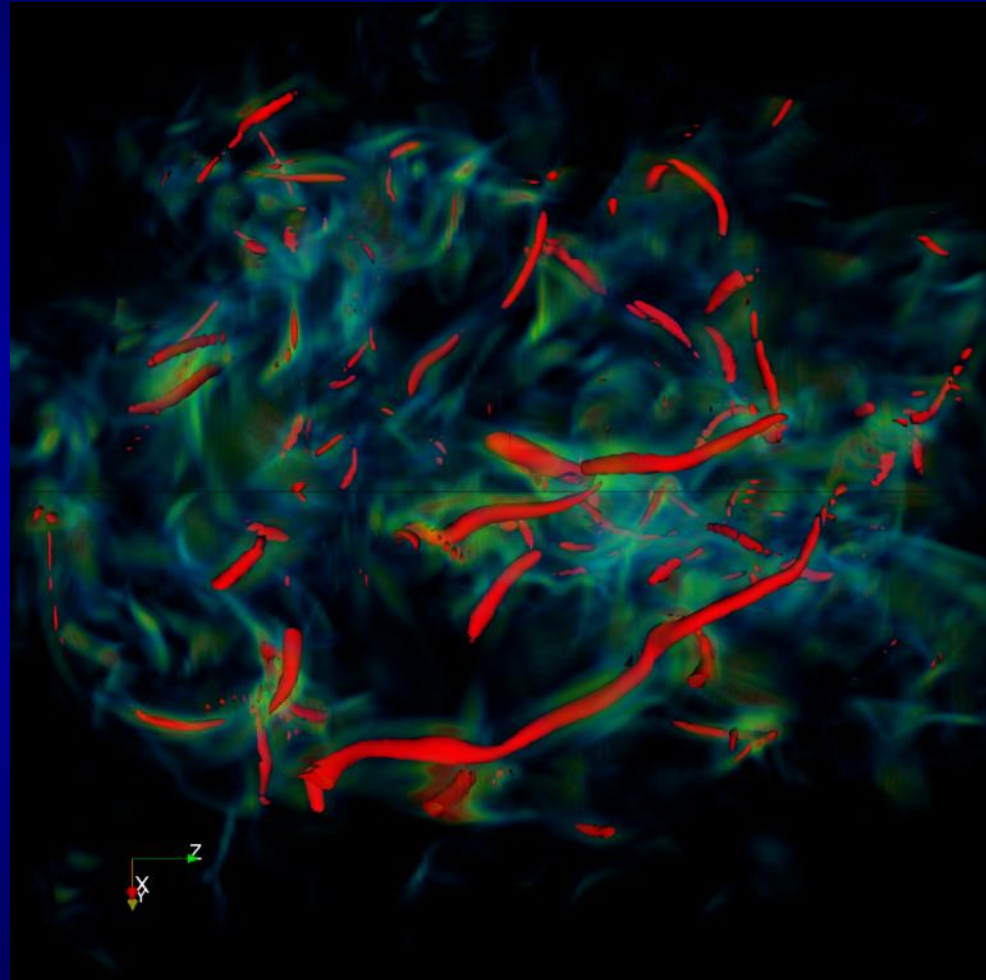


# Isosurfaces (Contours)

Plot the surface for a given scalar value.

Good for showing known values of interest

Good for sampling through a data range

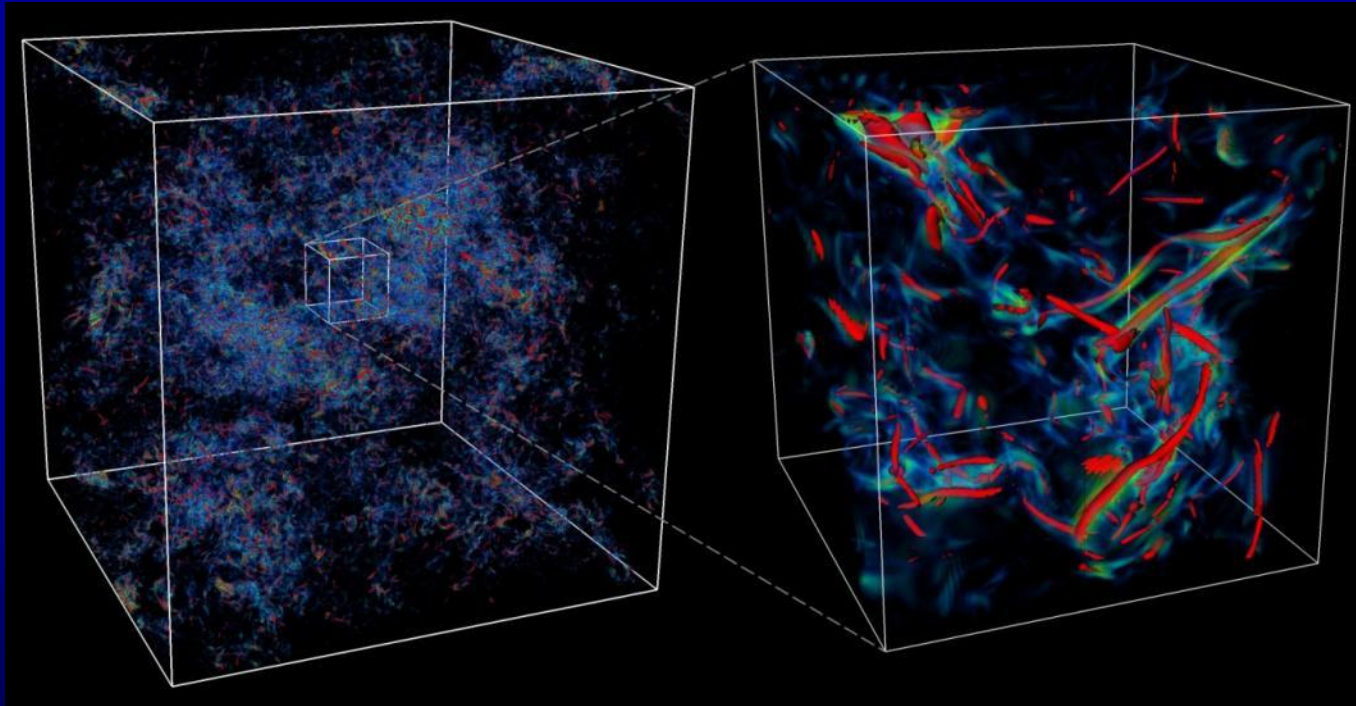


# Volume Rendering

Expresses how light travels through a volume

Color and opacity controlled by transfer function

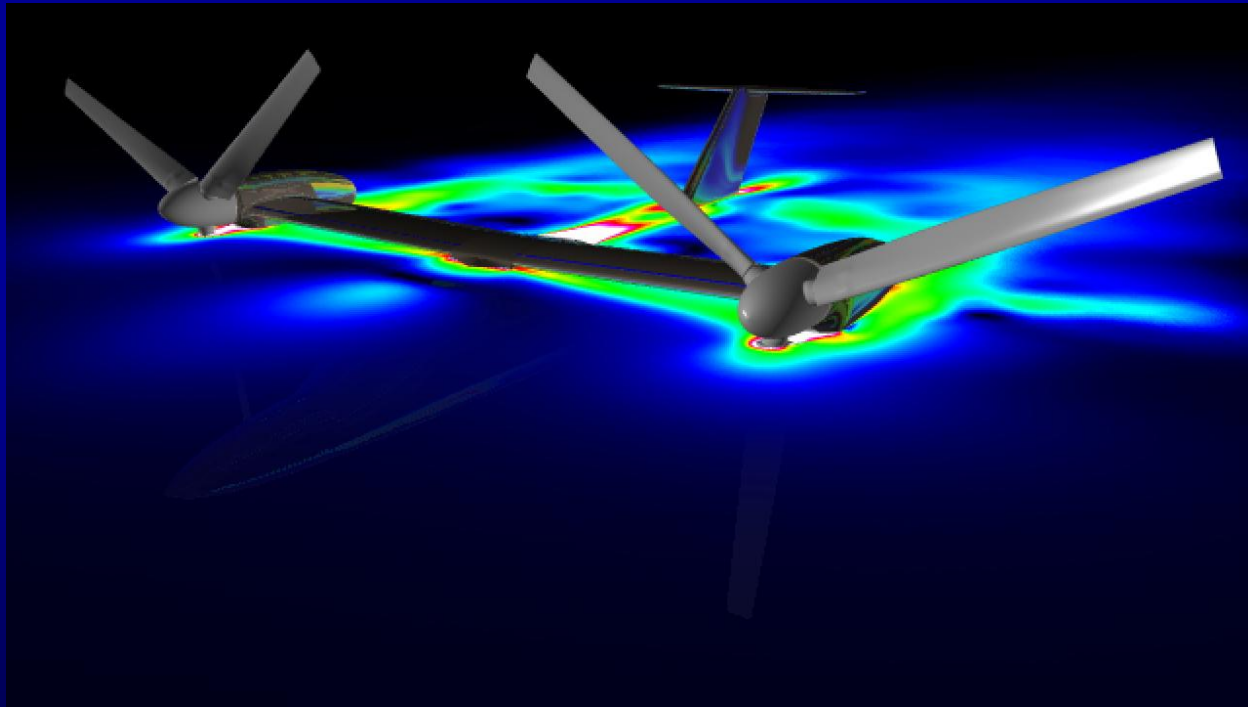
Smother transitions than isosurfaces



# Clipping / Slicing Planes

Extract a plane from the data to show features

Hide part of dataset to expose features



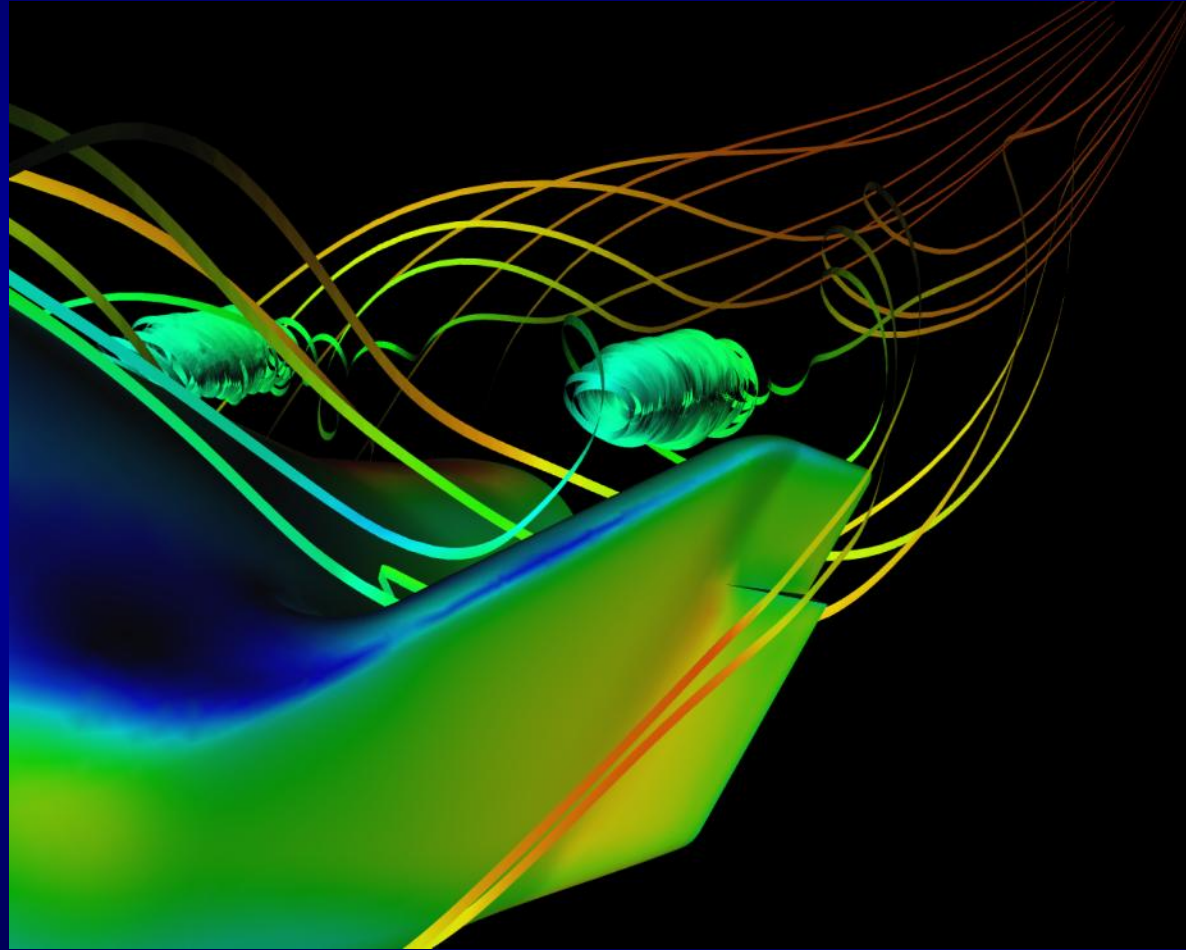
# Particle Traces (Streamlines)

Given a vector field, extract a trace that follows that trajectory defined by the vector.

$$P_{\text{new}} = P_{\text{current}} + V_P \Delta t$$

Streamlines – trace in space

Pathlines – trace in time



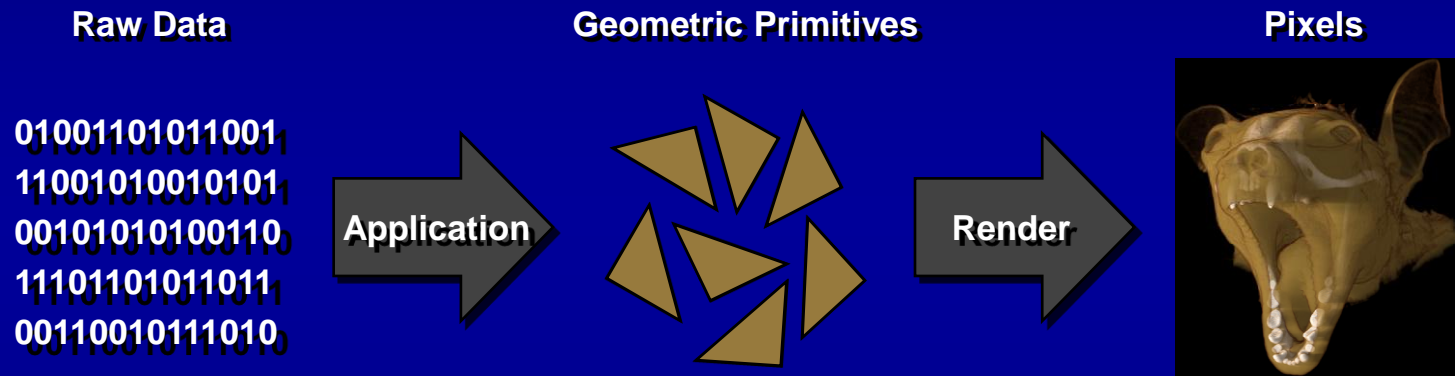
# Visualization Resources

- Personal machines
  - Most accessible, least powerful
- Projection systems
  - Seamless image, high purchase and maintenance costs
- Tiled-LCD displays
  - Lowest per-pixel costs, bezels divide image
- Remote visualization
  - Access to high-performance system, latency can affect user experience



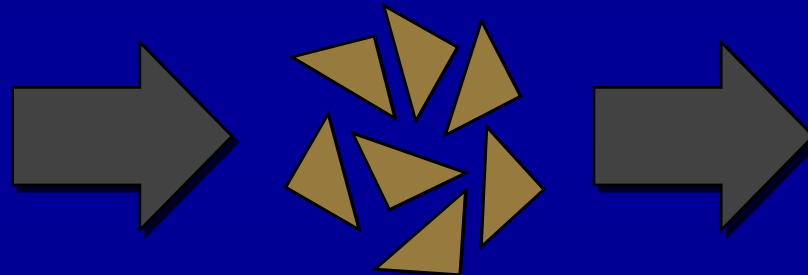
# Visualization Challenges

# Visualization Allows Us to “See” the Science



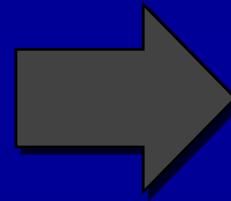
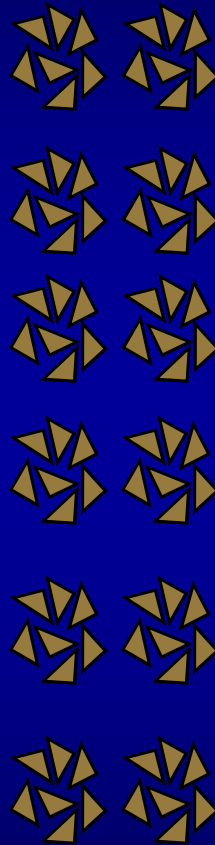
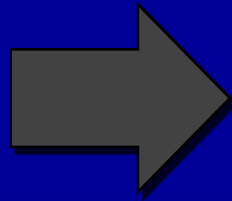
# But what about large, distributed data?

01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010  
01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010  
01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010  
01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010  
01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
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01001101011001 01001101011001  
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00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010



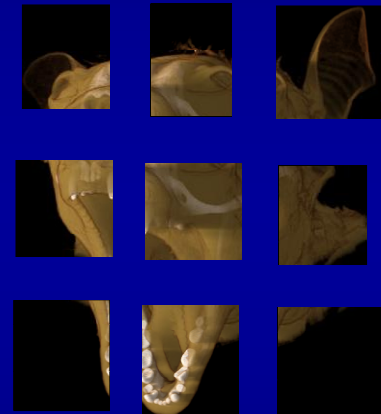
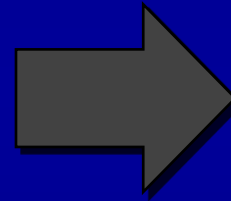
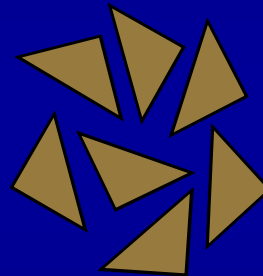
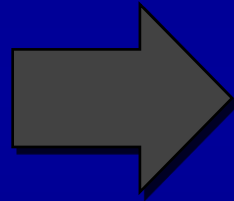
# Or distributed rendering?

01001101011001  
11001010010101  
00101010100110  
11101101011011  
00110010111010



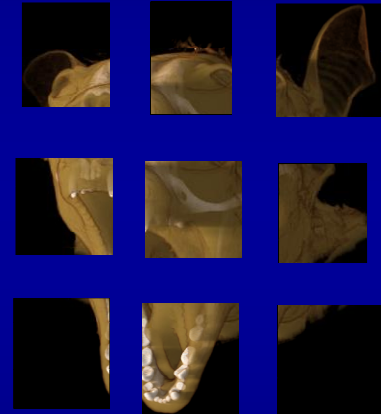
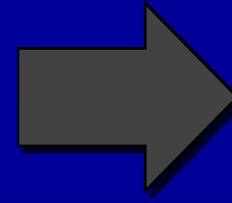
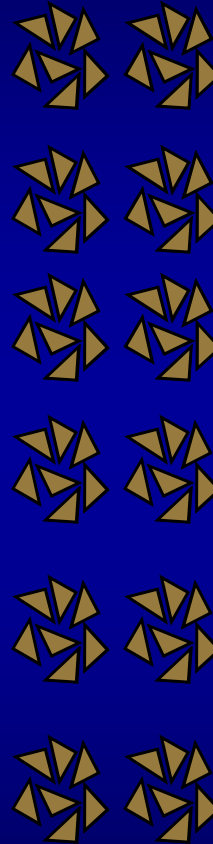
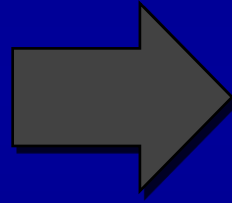
# Or distributed displays?

01001101011001  
11001010010101  
00101010100110  
11101101011011  
00110010111010



# Or all three?

01001101011001 01001101011001  
11001010010101 11001010010101  
00101010100110 00101010100110  
11101101011011 11101101011011  
00110010111010 00110010111010  
01001101011001 01001101011001  
11001010010101 11001010010101  
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00110010111010 00110010111010

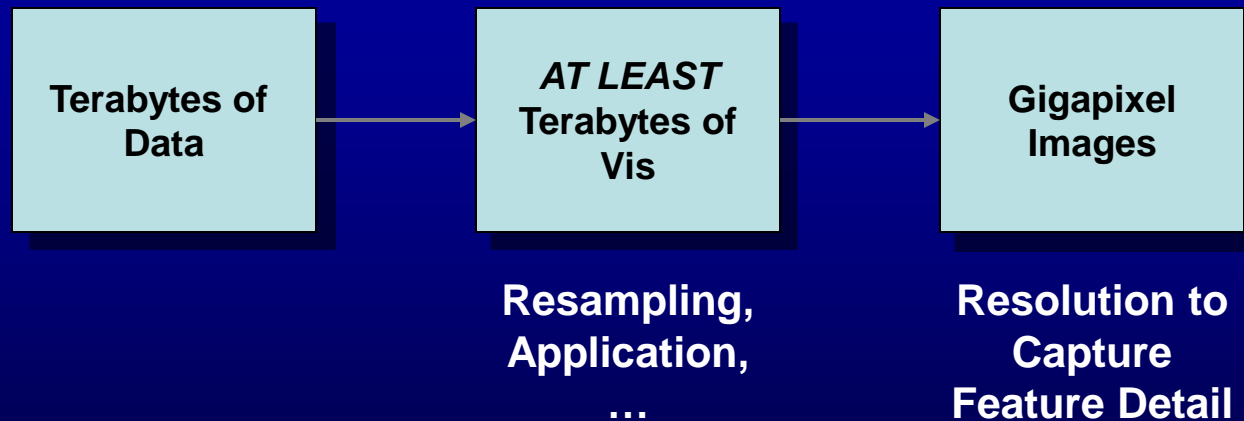


# Visualization Scaling Challenges

- Moving data to the visualization machine
- Most applications built for shared memory machines, not distributed clusters
- Image resolution limits in some software cannot capture feature details
- Displays cannot show **entire** high-resolution images **at their native resolution**

# Visualization scales with HPC

Large data produced by large simulations require large visualization machines and produce large visualization results





# Moving Data

- How much time do you have?

File Size	10 Gbps	54 Mbps
1 GB	1 sec	2.5 min
1 TB	~17 min	~43 hours
1 PB	~12 days	~5 years

# Analyzing Data

- Visualization programs **only beginning** to **efficiently** handle ultrascale data
  - 650 GB dataset -> 3 TB memory footprint
  - Allocate HPC nodes for RAM not cores
  - N-1 idle processors per node!
- Stability across many distributed nodes
  - Rendering clusters typically number  $N \leq 64$
  - Data must be dividable onto N cores

*Remember this when resampling!*

# Solution by Partial Sums

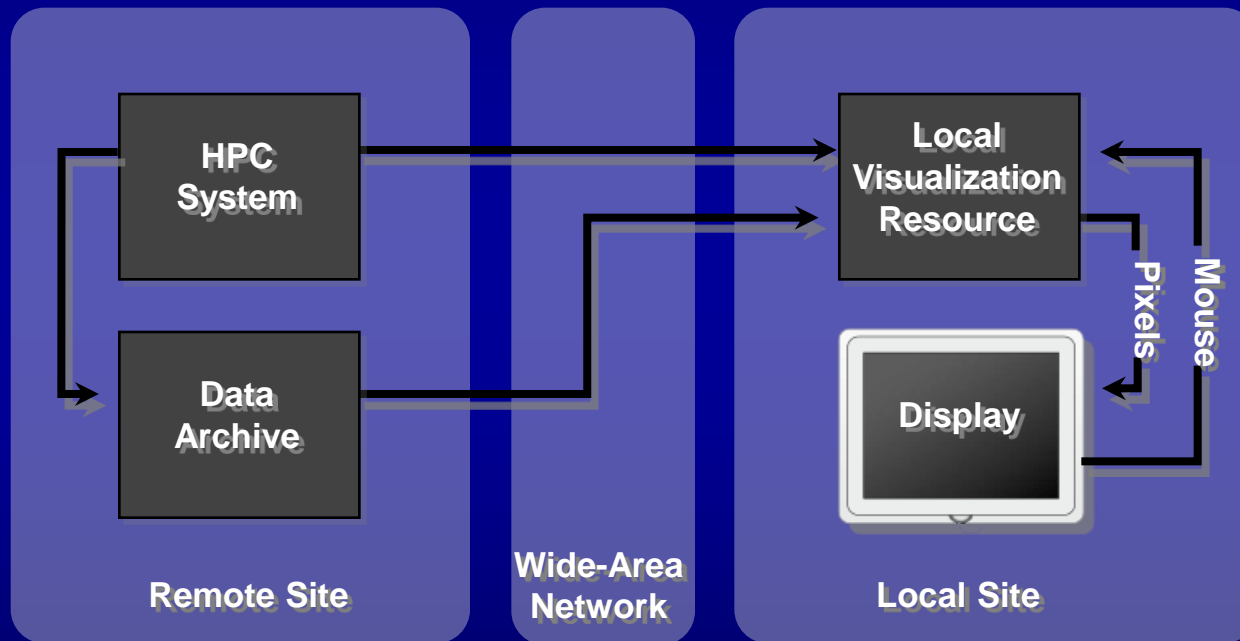
- Moving data – integrate vis machine into simulation machine. **Move the machine** to data!
  - Ranger + Spur: shared file system and interconnect
- Analyzing data – create larger vis machines and develop more efficient vis apps
  - Smaller memory footprint
  - More stable across many distributed nodes

Until then, **the simulation machine is the vis machine!**

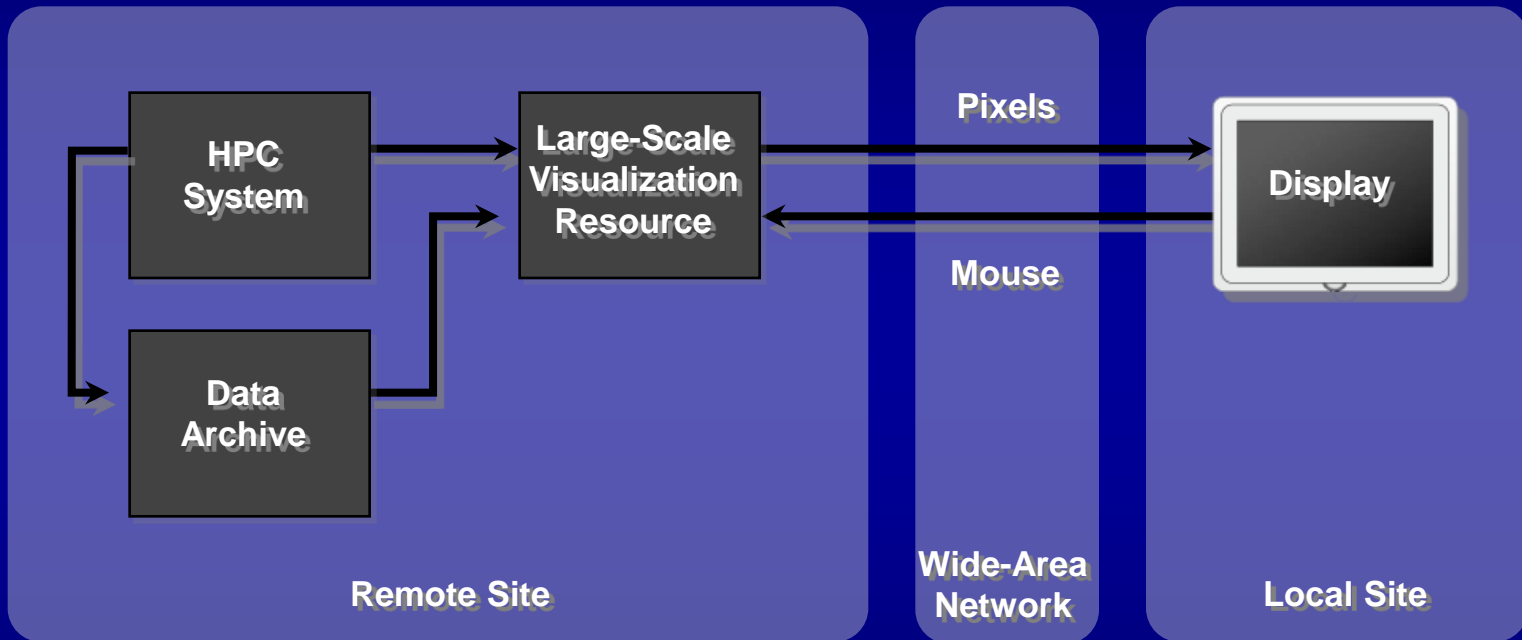
# Solution by Partial Sums

- Imaging data – focus vis effort on interesting features  
parallelize image creation
  - Feature detection to determine visualization targets  
**but can miss “unknown unknowns”**
  - Distribute image rendering across cluster
- Displaying data – high resolution displays  
multi-resolution image navigation
  - Large displays need large spaces
  - Physical navigation of display provides better insights

# Old Model (No Remote Capability)



# New Model Remote Capability



# Parallel Visualization

- Task parallelism – passing results to 1 process for rendering

**Timesteps**

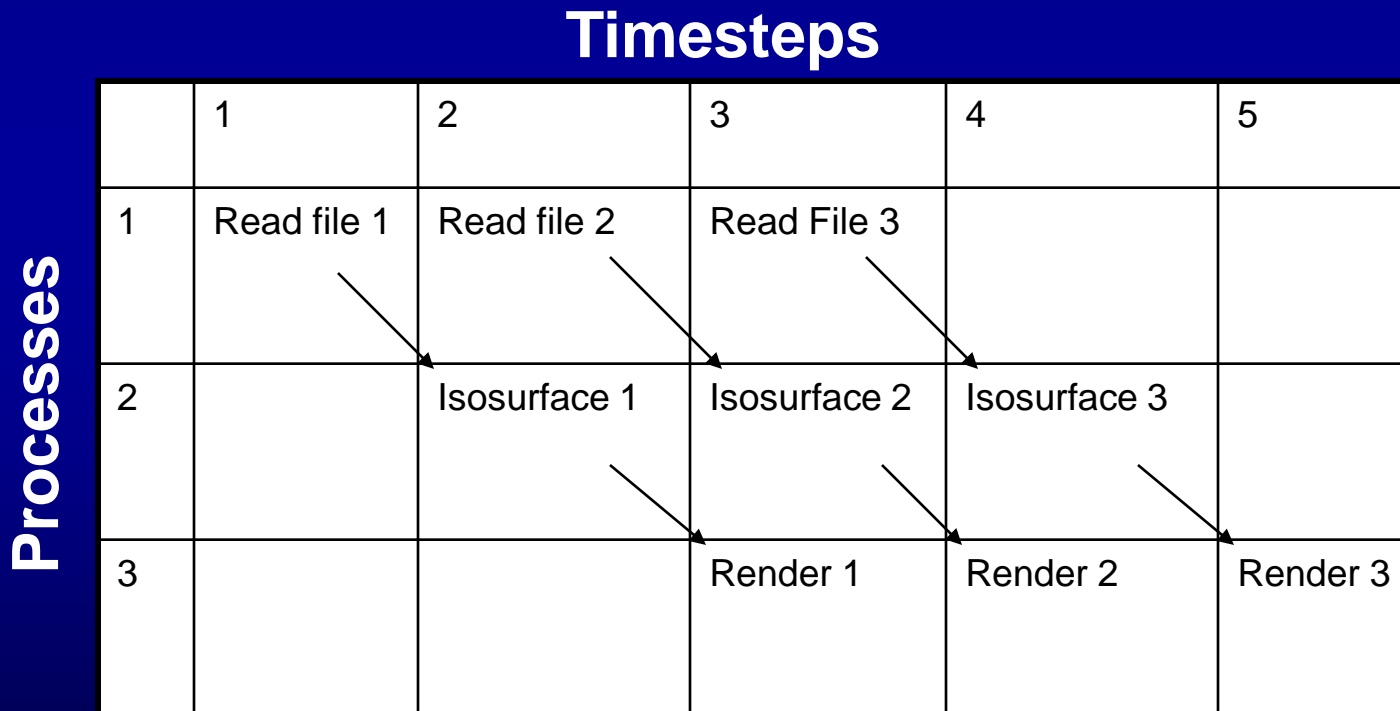
**Processes**

	1	2	3	4	5
1		Read file 1	Isosurface 1	Cut Plane 1	
2			Read file 2	Streamlines 2	Render
3	Read file 3	Triangulate 3	Decimate 3	Glyph 3	

# Parallel Visualization

## Pipeline parallelism

Useful when processes have access to separate resources or when an operation requires many steps.





# Parallel Visualization

## Data parallelism

Data set is partitioned among the processes and all processes execute same operations on the data.

Scales well **as long as the data and operations can be decomposed.**

		Timesteps		
		1	2	3
Processes	1	Read partition 1	Isosurface partition 1	Render partition 1
	2	Read partition 2	Isosurface partition 2	Render partition 2
	3	Read partition 3	Isosurface partition 2	Render partition 3

# Parallel Visualization Libraries

- Chromium – <http://chromium.sourceforge.net>
  - Sits between application and native OpenGL
  - Intercepts OpenGL calls, distribute across cluster
  - Can do either sort-first or sort-last (sort-first is simpler, sort-last can be better for large data)
  - Last update 31 Aug 2006, no new GL goodies
- IceT – <http://www.cs.unm.edu/~kmorel/IceT/>  
SAGE – <http://www.evl.uic.edu/cavern/sage/>  
CGLX – <http://vis.ucsd.edu/~cglx/>
  - specifically for large tiled displays
  - Must use IceT / SAGE / CGLX API in code
- Mesa – <http://www.mesa3d.org/>
  - Software rendering library
  - Enables OpenGL rendering on machines without GPUs

# Open-Source Parallel Vis Apps

- VisIt – <https://wci.llnl.gov/codes/visit/>
  - Good scaling to hundreds of cores
  - Integrated job launching mechanism for rendering engines
  - Good documentation and user community
  - GUI not as polished
- ParaView – <http://www.paraview.org/>
  - Polished GUI, easier to navigate
  - Less stable across hundreds of cores
  - Official documentation must be purchased, though rich knowledge base on web (via Google)

# CUDA – coding for GPUs

- C / C++ interface plus GPU-based extensions
- Can use both for accelerating visualization operations and for general-purpose computing (GPGPU)
- Special GPU libraries for math, FFT, BLAS

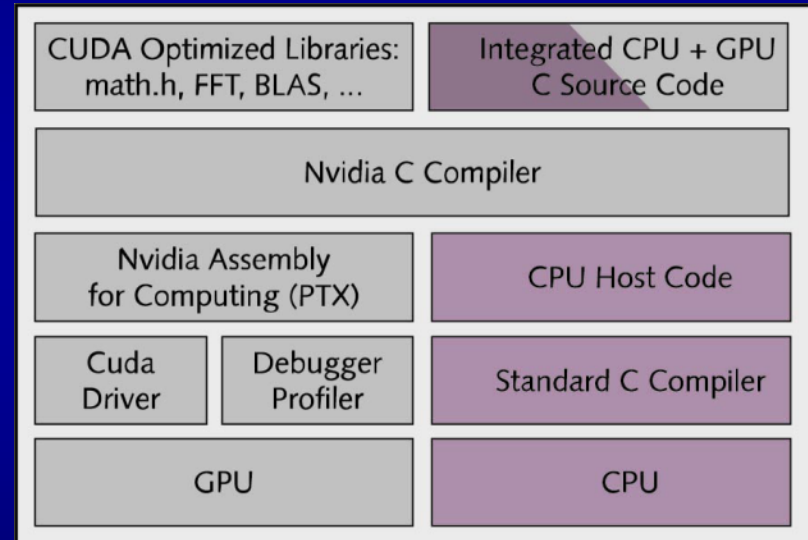


Image: Tom Halfhill, Microprocessor Report

# GPU layout

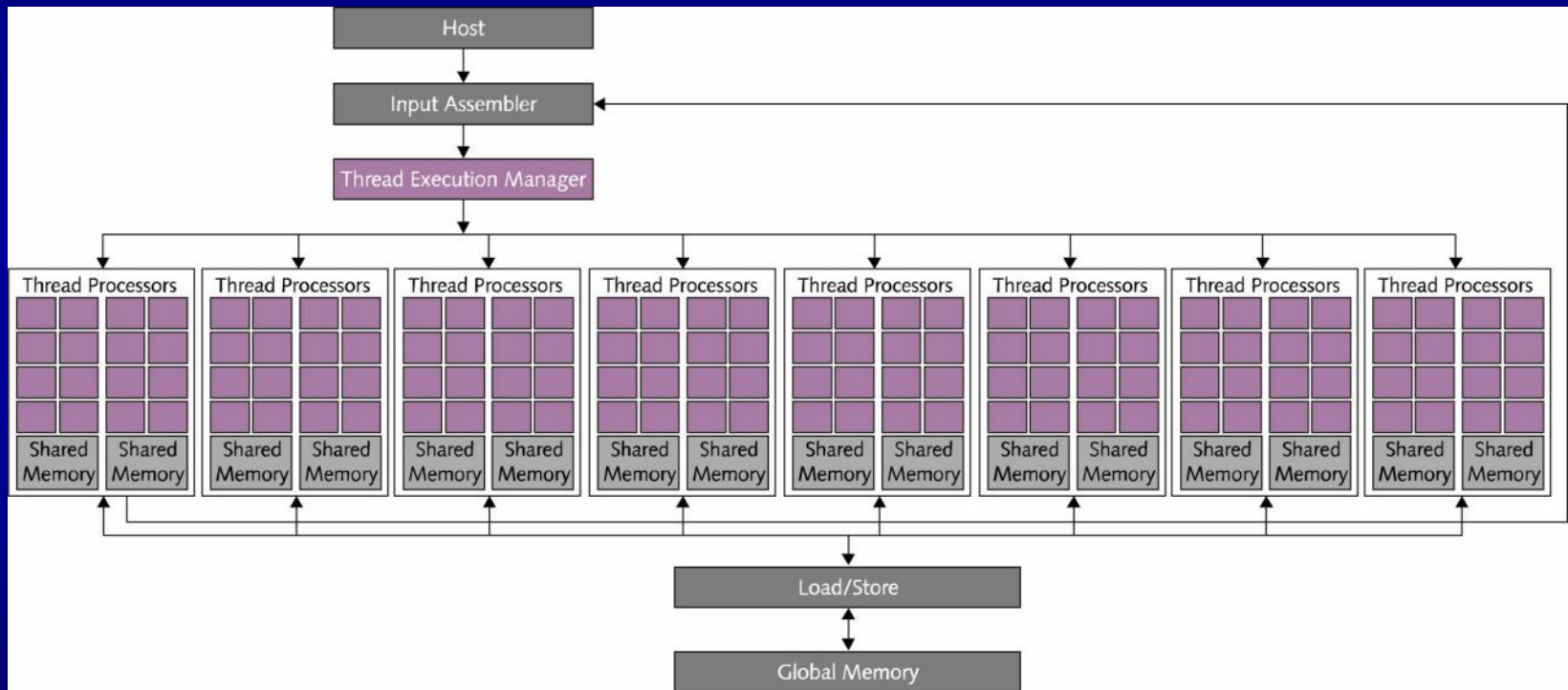


Image: Tom Halfhill, Microprocessor Report

# GPU Considerations

- Parallelism – kernel should be highly SIMD
  - Switching kernels is expensive!
- Job size – high workload per thread
  - amortize thread initialization and memory transfer costs
- Memory footprint – task must decompose well
  - local store per GPU core is low (16 KB on G80)
  - card-local RAM is limited (~1GB on G8x)
  - access to system RAM is slow (treat like disk access)
- GPU N-body study (in GPU Gems 3):  
[http://www.nvidia.com/object/io\\_1195170003876.html](http://www.nvidia.com/object/io_1195170003876.html)

# Summary

- Challenges at every stage of visualization when operating on large data
- Partial solutions exist, though not integrated
- Problem sizes continue to grow at every stage
- Vis software community must keep pace with hardware innovations



Thank you!

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