



MTAGS11 – Seattle, Washington

Parallel High-Resolution Climate Data Analysis using Swift



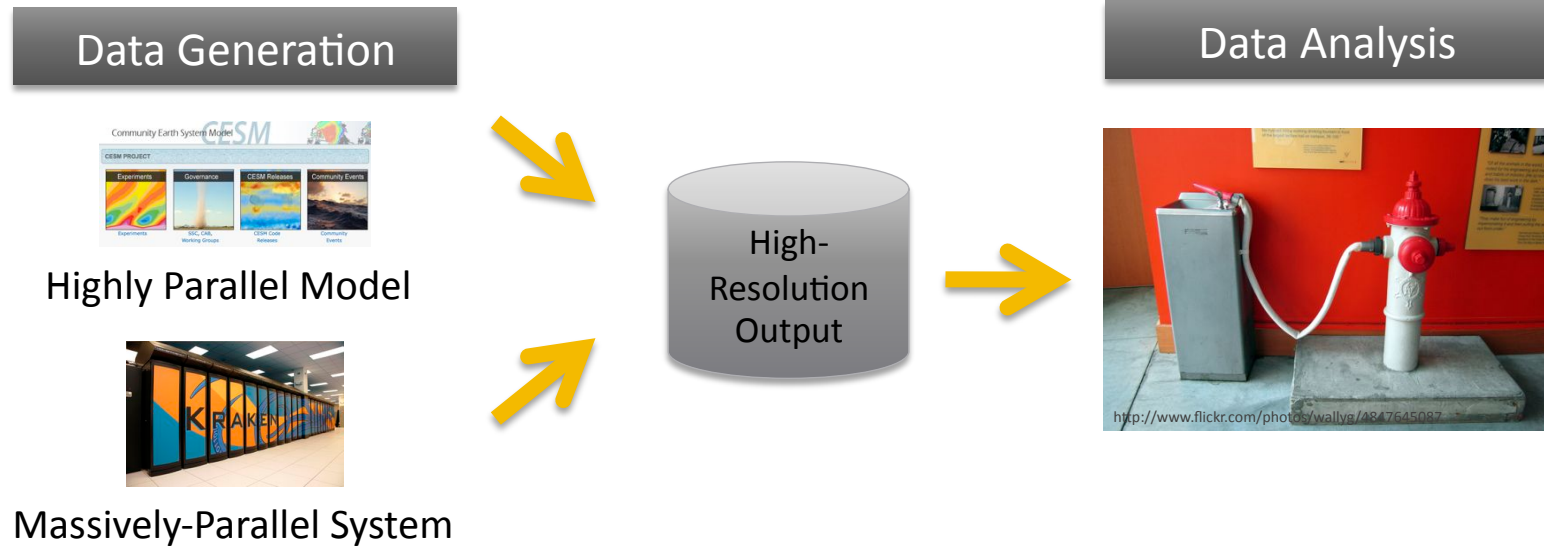
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National Center for Atmospheric Research

Overview: Many-Task & Data-Intensive



- Complimentary aspects
 - Workflow development: generation and analysis
 - Infrastructure support

Science: The PetaApps Project

- High-resolution climate experiment
 - Explore impact of weather noise on climate... and technical/computer science issues
 - ~18M CPU hours on NICS Kraken (np=5844)
- “Supersizing” the data: ~100TB for 155 years
 - 0.1° Ocean [3600 x 2400 x 42] } 100x
 - 0.1° Sea-ice [3600 x 2400 x 20] }
 - 0.5° Atmosphere [576 x 384 x 26] } 4x
 - 0.5° Land [576 x 384] }

Workflow: The AMWG Diagnostic

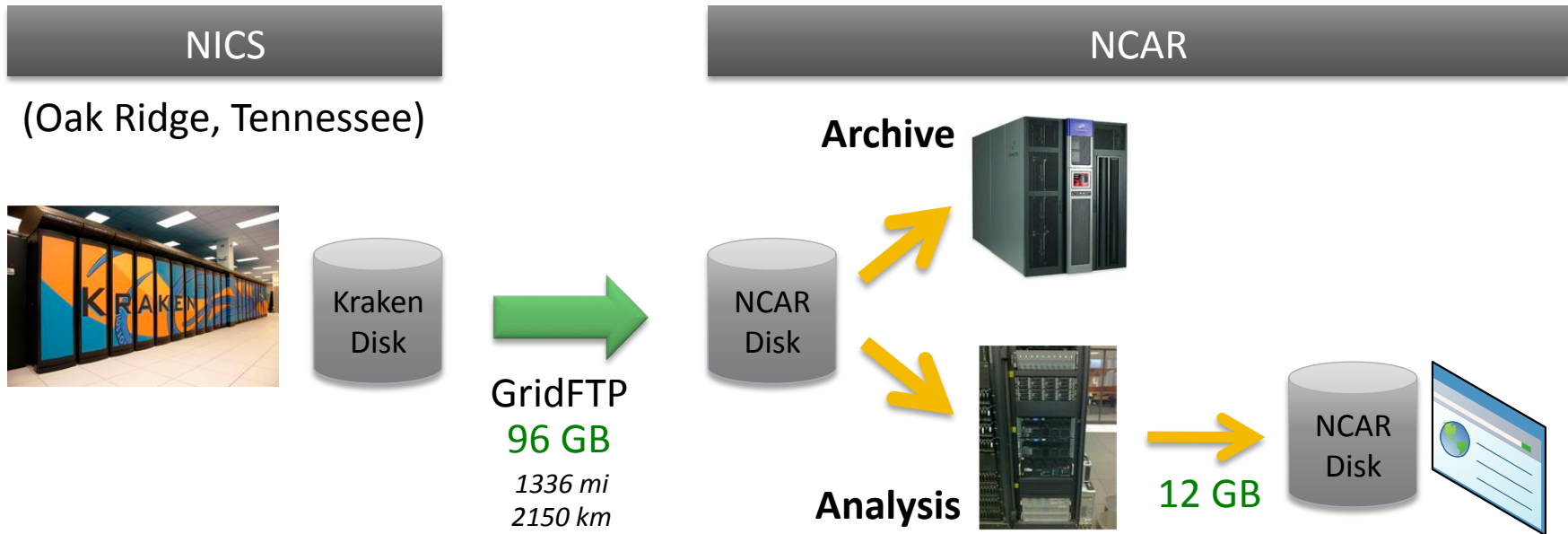
- Analysis process for CESM atmosphere component
- Hardware over the years
 - Past: Tape as a file server
 - Now: Central shared disk
 - Emerging: Data-intensive platforms
- Model constantly changing, resolution increasing, and hardware improving...
but it's the same C-shell script!



John
(Scientist)

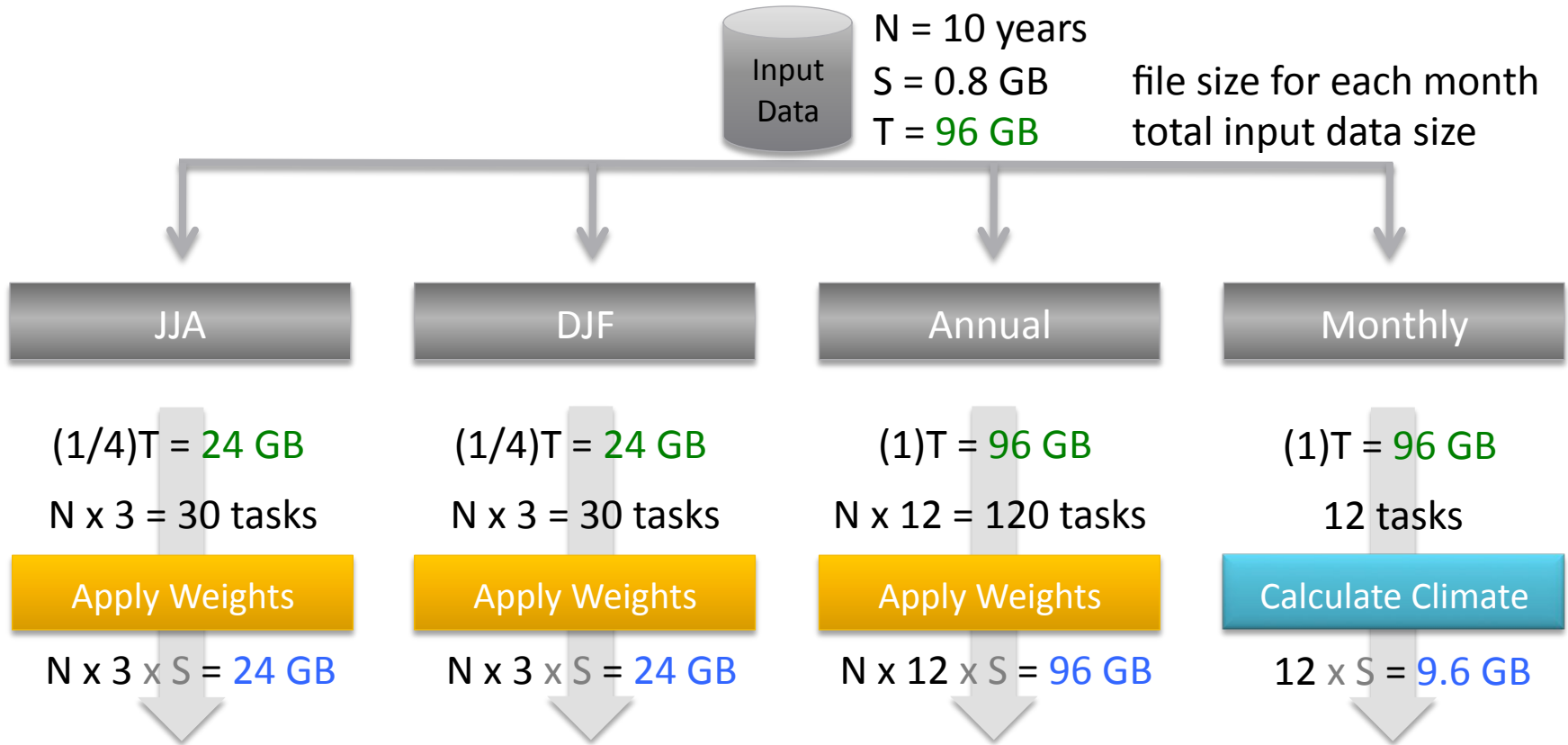
This hasn't really changed since 2001.

The AMWG Analysis Pipeline



- Original inefficiencies
 - Analysis “diagnostic” process is a **serial** C-shell script; 2° to 0.5° → *minutes to hours*
 - Major components invoked manually

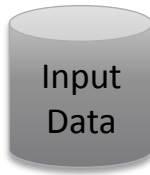
The AMWG Diagnostic Workflow

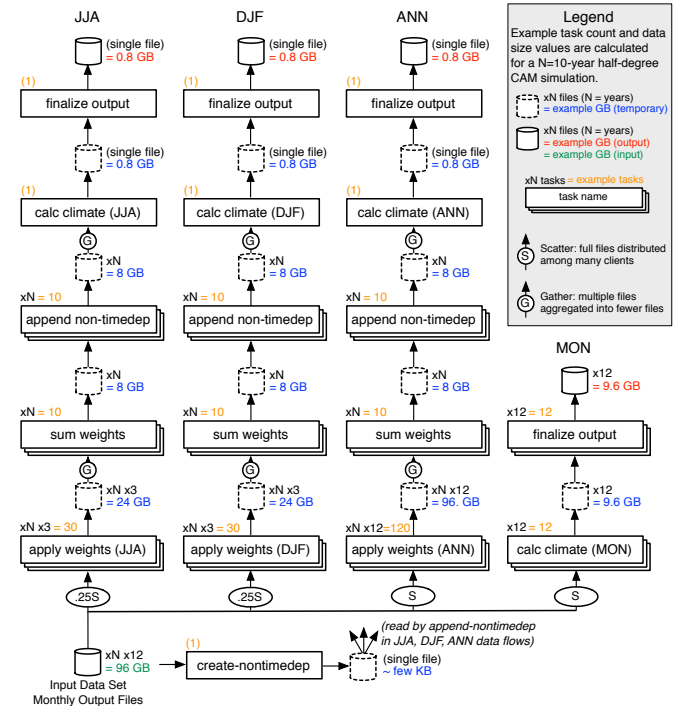


- First step: Given 96 GB of input data, read 240 GB and write 153.6 GB

The AMWG Diagnostic Workflow

- A data-intensive, many-task workflow


 10 year simulation
 120 files (one per month)
 0.8 GB file size
96 GB total input data size



- Workflow data handling volume

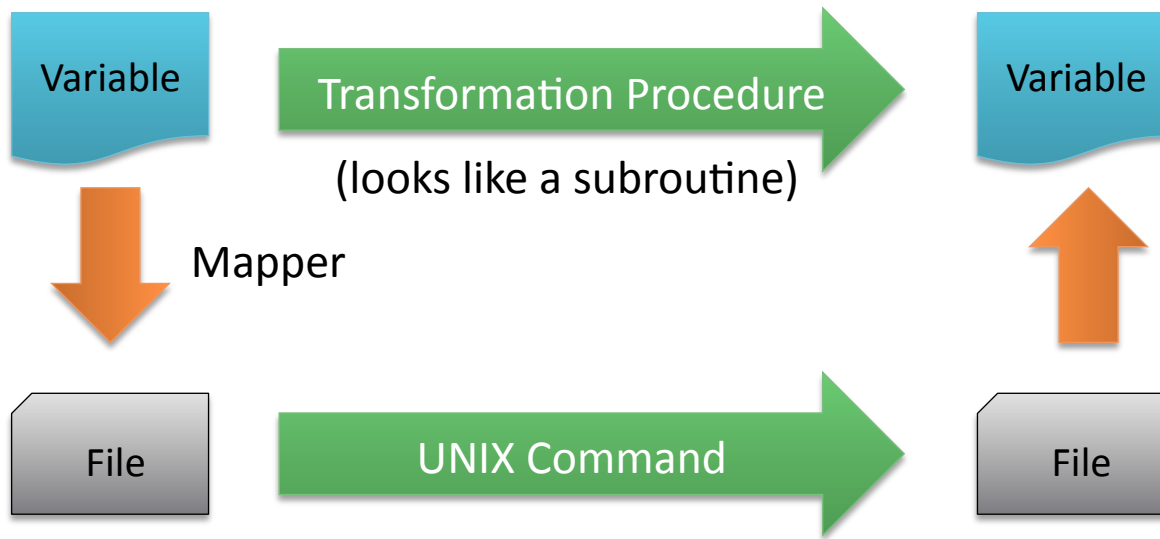
Input data	96 GB
Read from disk	444 GB

Output data	12 GB
Write to disk	194 GB

Swift Parallel Scripting

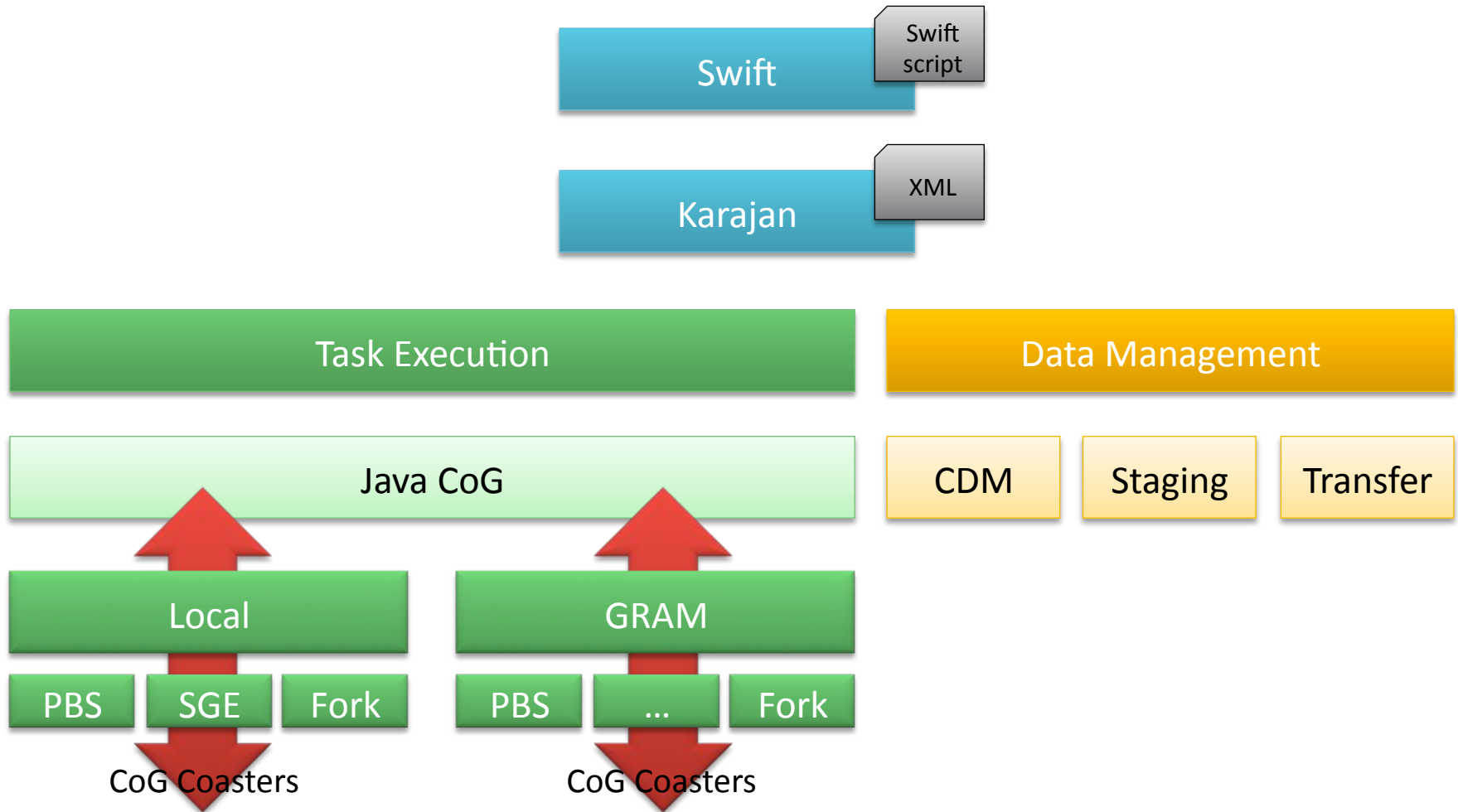


- A scripting language designed for **transforming data stored in files**



- **Trigger-based execution** exposes parallelism

Swift: Coordinating Tasks on Sites



Data Management

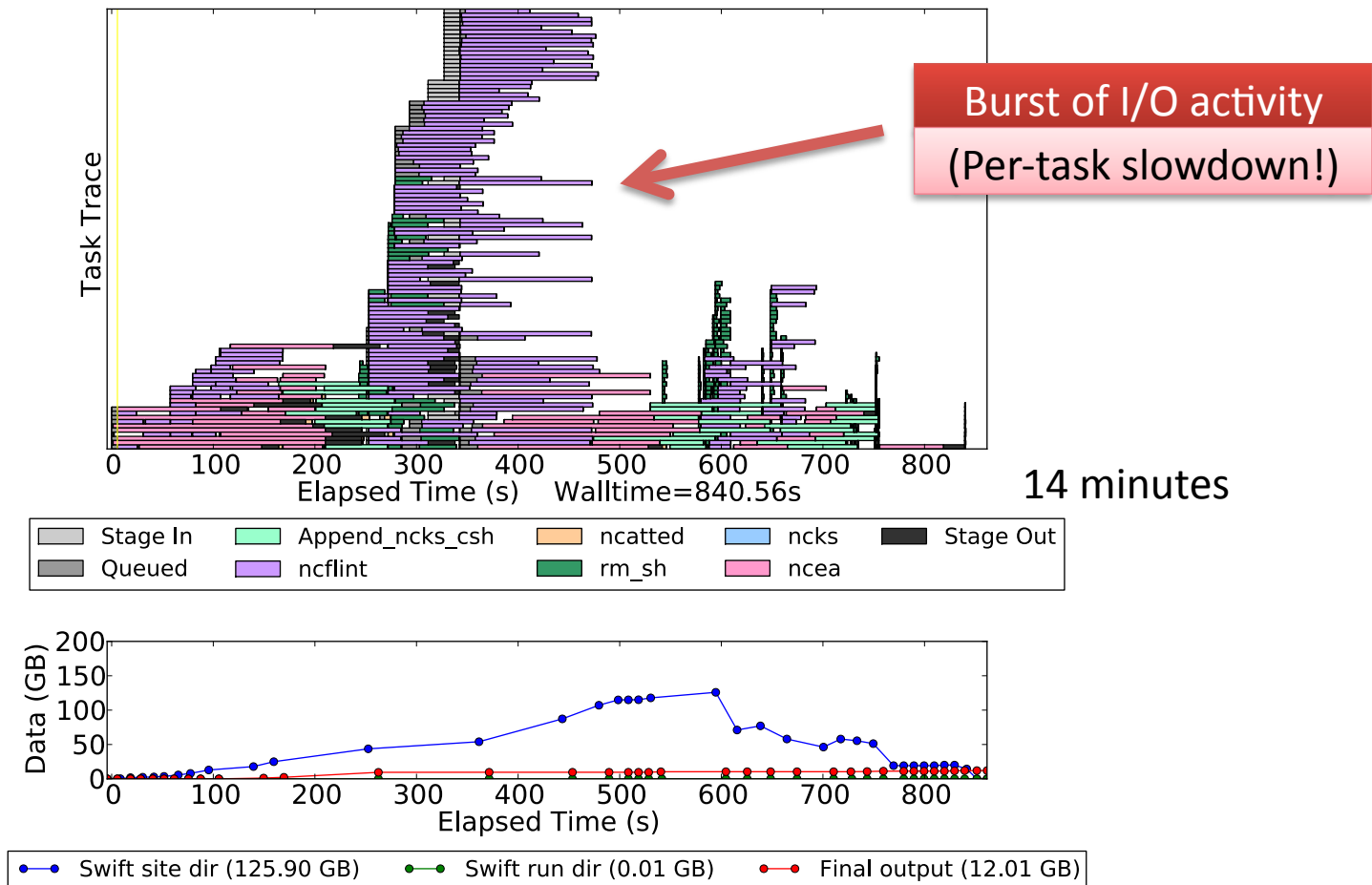
- Swift manages the runtime data environment
 - Distribute tasks across multiple sites and Grids
 - Ensure re-startability and task independence
- But it didn't match our single-site paradigm...



Data Management Controls

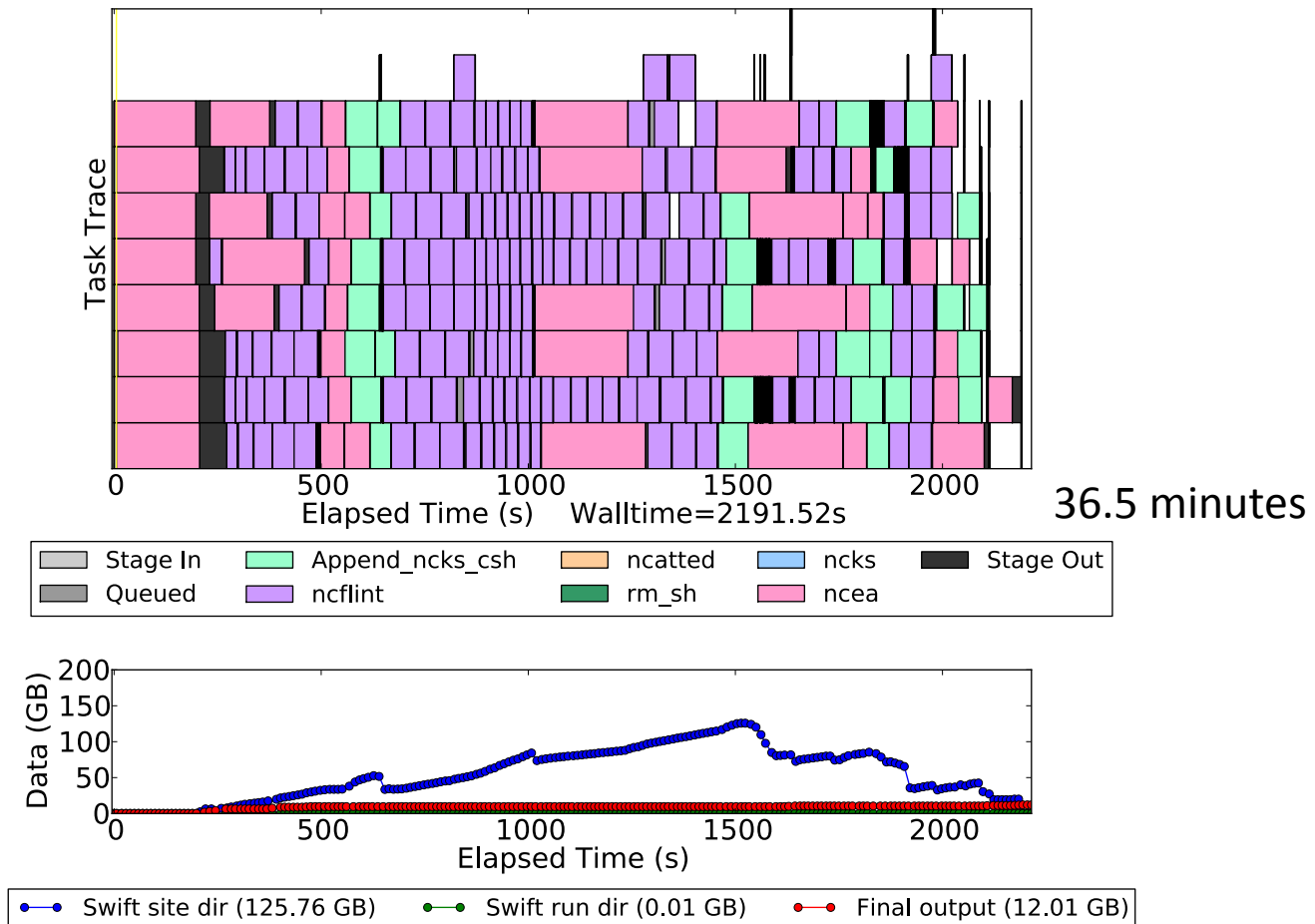
- Staging
 - Problem: Copy overhead, doubled capacity
 - Solution: Swift Collective Data Management (CDM)
- Retention
 - Problem: Capacity constraints and *intermediate* files
 - Our solution (about a year ago):
 - Artificial parallelism constraints on stages
 - Manual file management
 - Or: Swift data management using variable scope

Parallelism in the AMWG Diagnostic



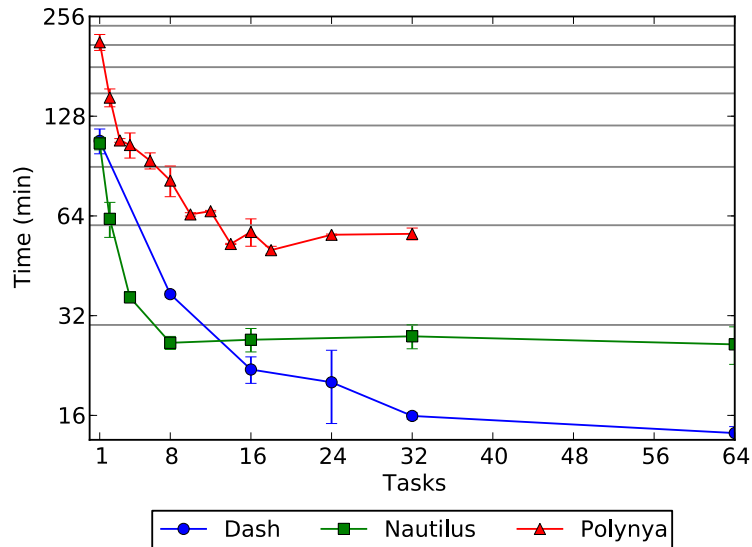
This version constrains temporary file generation by forcing serialization between the four analysis chains and removing earlier temporary files.

Parallelism in the AMWG Diagnostic



This run allows only 8 threads to be in an execution stage. Some operations, such as task clean-up, might overlap.

AMWG/Swift Prototype Performance



Comparing data analysis architectures

Dash - Linux cluster at SDSC with 32 nodes, 16 cores per/node, and 48 GB/node; GPFS-WAN storage (without ScaleMP)

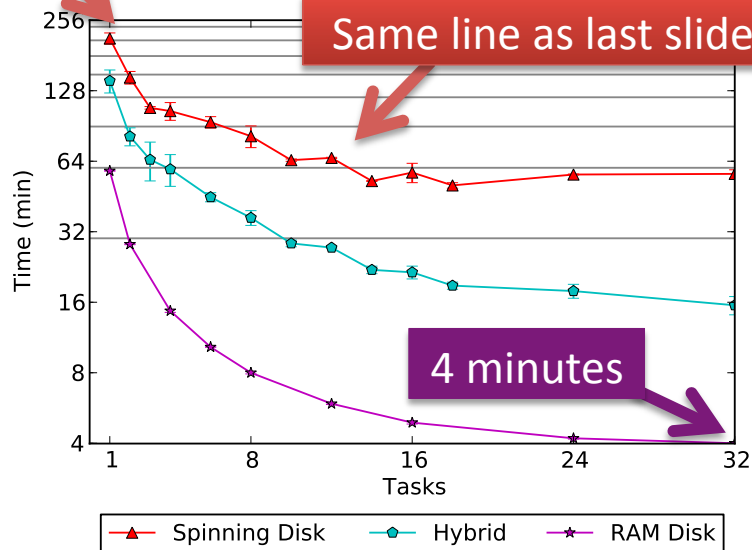
Nautilus - SGI Altrix UV 1000 at NICS with 4 GB/core (SSI); GPFS medusa Storage

Polynya - server at NCAR with 32 cores and 1 TB RAM, 2007-era GPFS from Frost

- Performance factors:
 - Limited workflow parallelism
 - Platform characteristics (I/O throughput)

AMWG/Swift Prototype Performance

214 minutes serial



Comparing storage technologies on Polynya

Spinning Disk – input and temporary data on parallel file system disk (**56 min**)

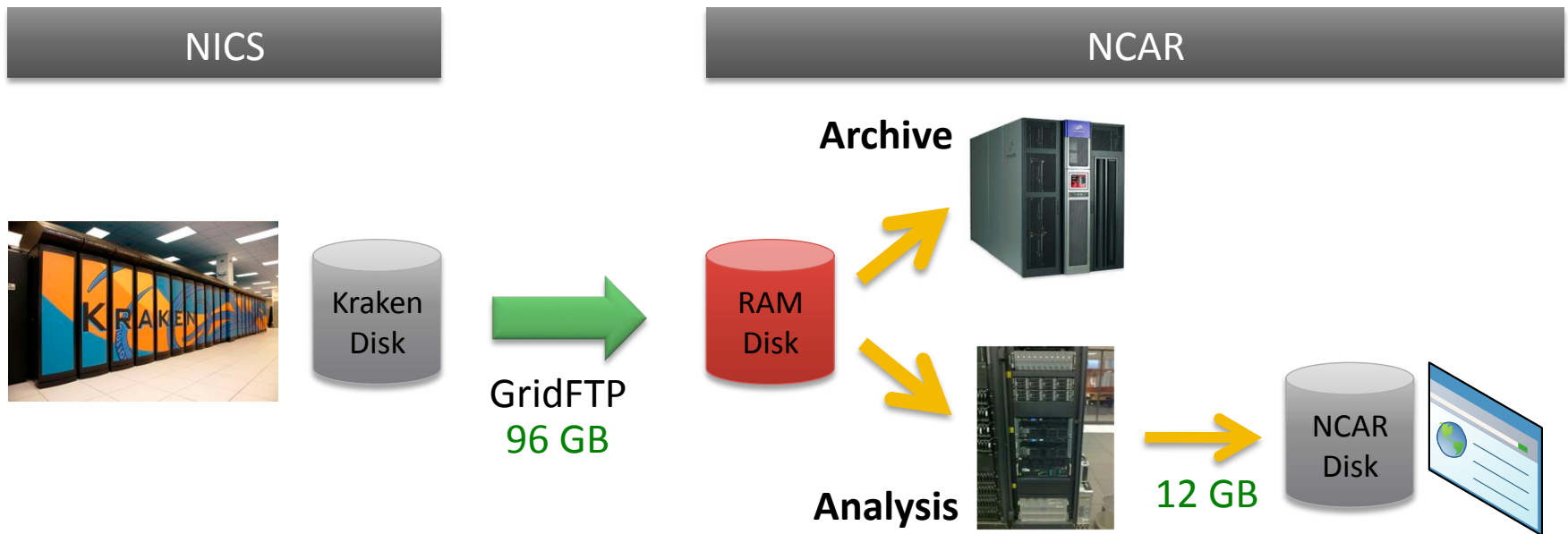
Hybrid – input on disk, temporary data on RAM disk (realistic scenario, **16 min**)

RAM disk – input and temporary data on RAM disk (pre-staged input, **4 min**)

- RAM disk provides substantial speedup
 - But isn't pre-staging the data *cheating*?
 - We also use the data multiple times...

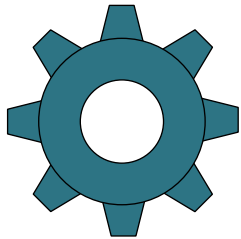
Current Work: Skipping Spinning Disk

What if we skipped the NCAR disk?



“Pre-staging” can be production mode!

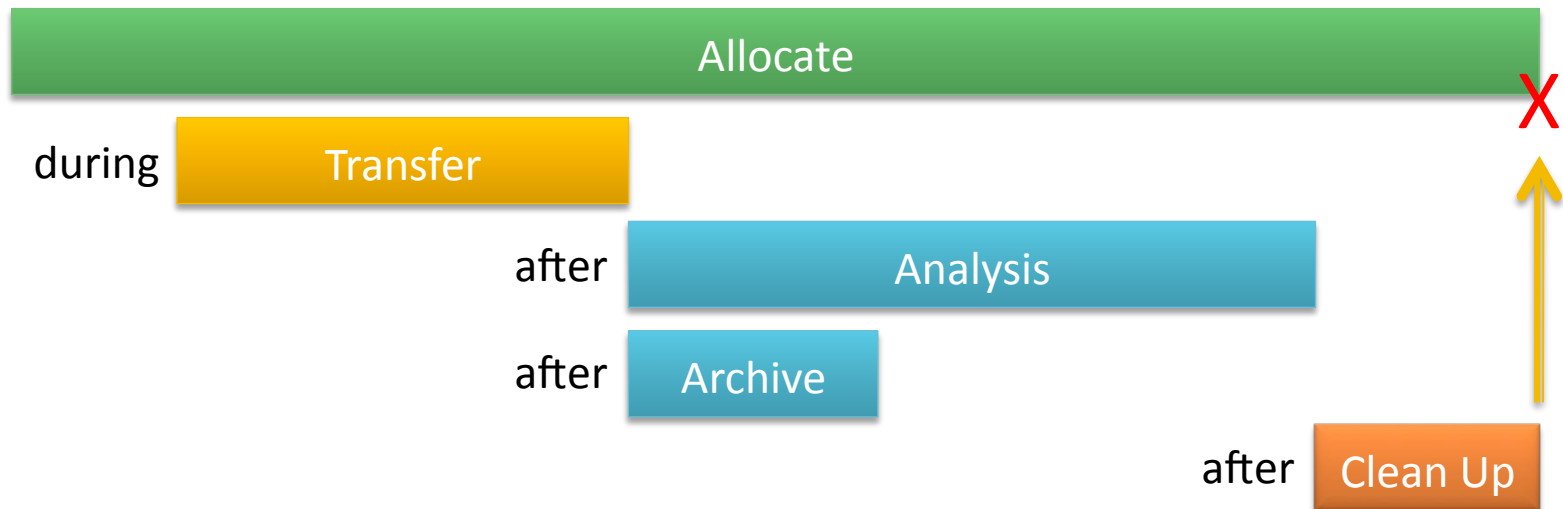
Vision: Automated Analysis Pipeline



- Simulation finishes on Kraken
- Kraken job triggers pipeline on Polynya using secure ssh key
- Automated pipeline:
 - Allocates storage space
 - Retrieves data from Kraken
 - Parallel jobs for:
 - Archive to tape (from RAM disk!)
 - Analysis scripts (using RAM disk!)
 - Cleans up and notifies human

Implementation: Scheduler Support

- **Co-scheduling** data and computation
 - Scheduling *data storage* on shared RAM disk
 - Scheduling *data transfers* with scheduled storage



- **Departure from traditional quota paradigm**

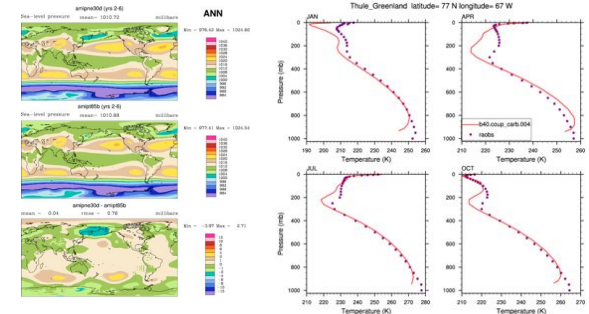
Future Work: First, the Bigger Fish

- This exploratory project:
 - Motivated by NSF PetaApps
 - Swift (2010), systems (2011)

- **ParVis** collaborative project:

- PI: Robert Jacob, ANL
- Argonne, Sandia, PNNL, NCAR, UC-Davis
- Broad-spectrum approach: PnetCDF, Swift, cloud computing, compression, NCL, data transfer

Parvis philosophy: Insight about climate comes mostly from computationally undemanding (to plot) 2D and 1D figures.

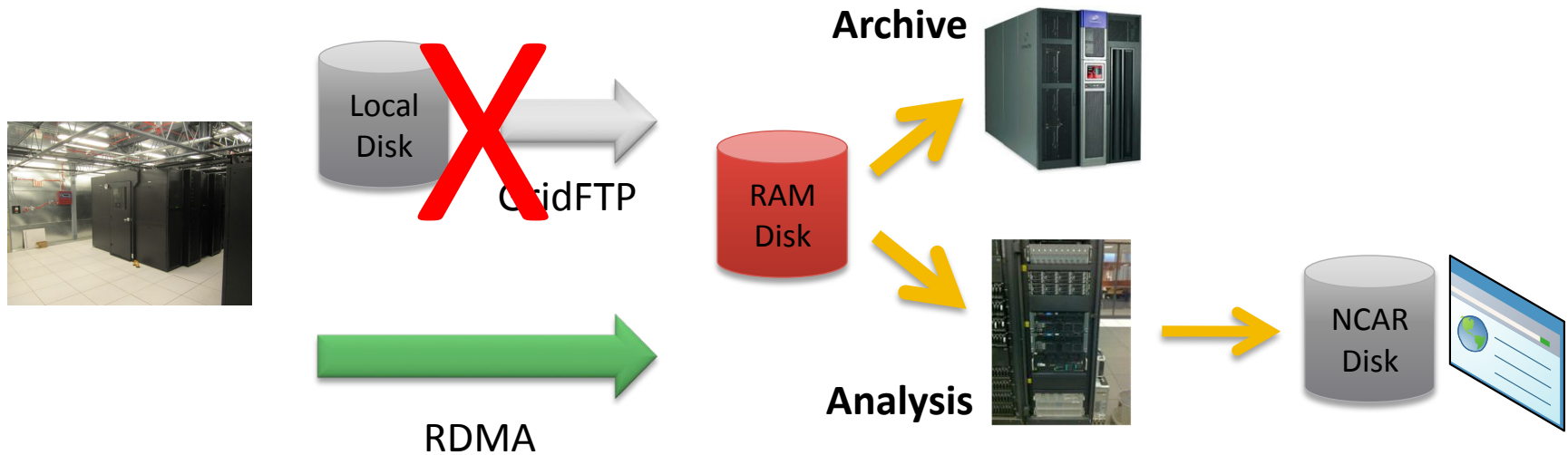


Why? The atmosphere and ocean have a small aspect ratio; 10,000 km vs. 10 km.

<http://www.cesm.ucar.edu/events/ws.2011/Presentations/Software/jacob.pdf>

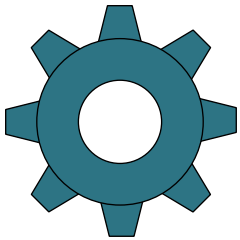
Future Work: Data Direct Deposit

What if we skipped all of the disk?

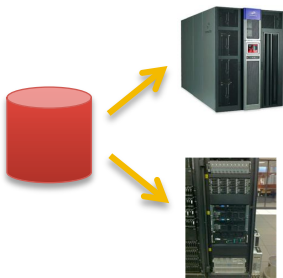
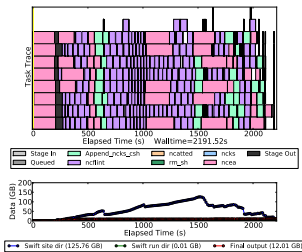


- New networking CI grant (NSF SDCI)
- Systems, networks, security... **at scale**

Summary and Conclusions



- AMWG workflow
 - Many-task and data-intensive
- Swift workflow management
 - Easily applied to this workflow
 - Applies to a wide range of platforms
 - Performance and reasonable scalability
- Future: a systems-centric solution including hardware and software



Acknowledgements

- Application and Workflow

- Taleena Sines, Michael Arndt, John Dennis

I need more storage!



John
(Scientist)

- Systems

- Dmitry Duplakin, Allan Espinosa
- Michael Oberg, Haiying Xu

I need better quota enforcement!



Oberg
(Engineer)

- Material and assistance from my colleagues

- Guy Cobb, Paul Marshall, Theron Voran, Rich Loft, Henry Tufo

- Other Centers and the PetaApps Project

- Chicago CI: Michael Wilde, Justin Wozniak, Mihael Hategan
- SDSC Dash: Allan Snavely, Shawn Strande, Adam Jundt, Jeffrey Bennett, Eva Hocks
- PetaApps: Kinter (COLA), Kirtman (Miami), Yelick (Berkeley), [Dennis et. al. \(NCAR\)](#), Bitz (Washington), and many more



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