

Science Impact of Sustained Cyberinfrastructure: The Pegasus Example

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University of Southern California,
Information Sciences Institute

NSF Office of Advanced CyberInfrastructure Webinar

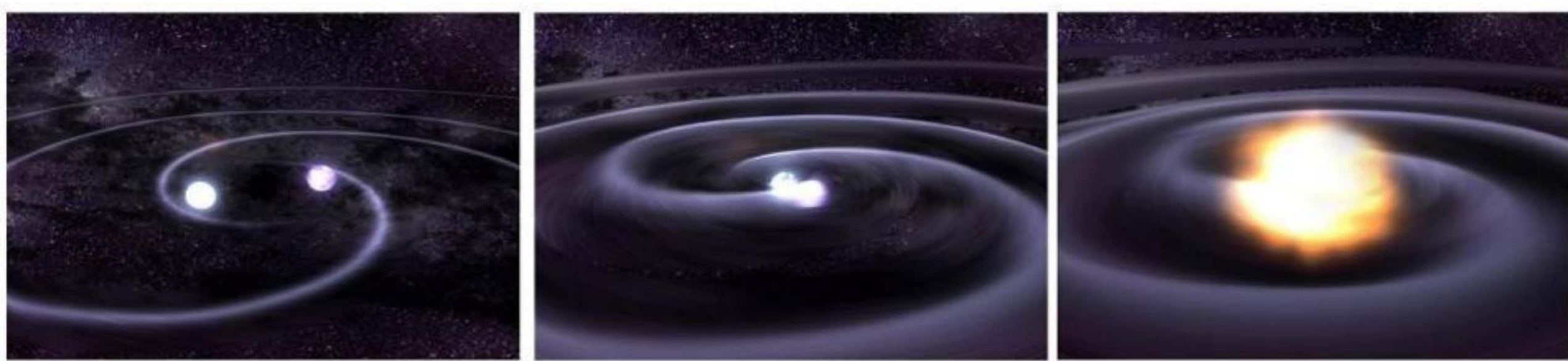
May 17, 2018

Pegasus represents a
long standing
collaboration with
Miron Livny,
University of
Wisconsin, Madison



October 16th 2017: “ LIGO and Virgo make first detection of gravitational waves produced by colliding neutron stars”

And kick off a new era of multi-messenger astronomy

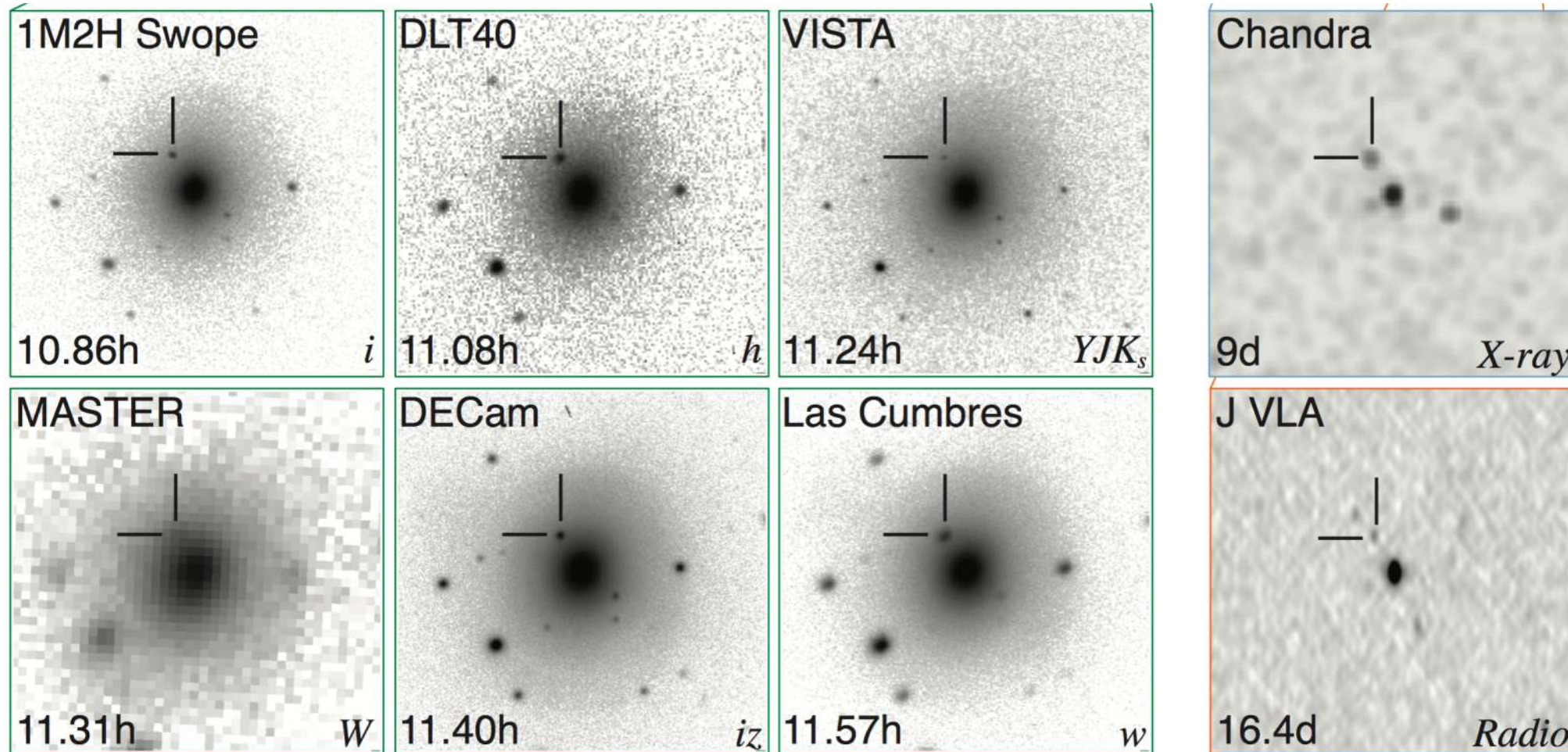


“The inspiral and merger of two neutron stars, as illustrated here, should produce a very specific gravitational wave signal, but the moment of the merger should also produce electromagnetic radiation that's unique and identifiable as such.”, credit LIGO

NASA's Fermi space telescope had detected a burst of gamma rays at about the same time



Targeting Telescopes on the Neutron Star Merger



“aftermath of the BNS merger.. On the left are six optical images taken between 10 and 12 hours after the merger by different telescopes. On the right are images constructed from x-ray and radio observations. The x-ray image was taken 9 days after the merger by NASA's Chandra X-ray Observatory. 16 days after the merger NRAO's Jansky Very Large Array (VLA) captured the radio image” from LIGO.org



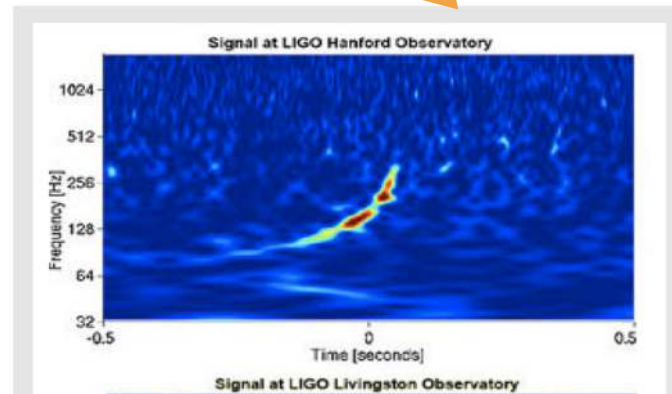
Dependable Cyberinfrastructure

Takes time, the Pegasus and LIGO Partnership

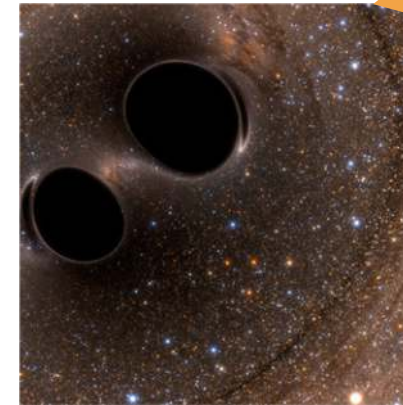
Nobel Prize



First Pegasus prototype



Blind injection detection



First detection of a GW from black hole collision

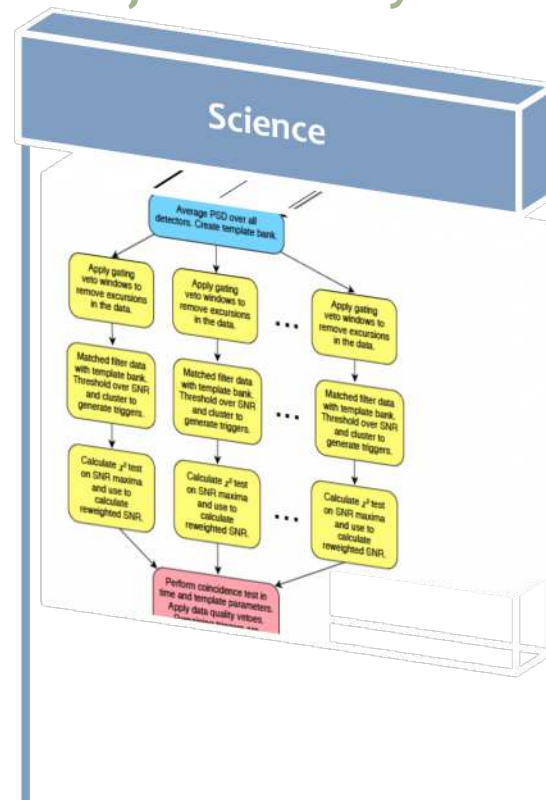


Multi-messenger neutron star merger observation

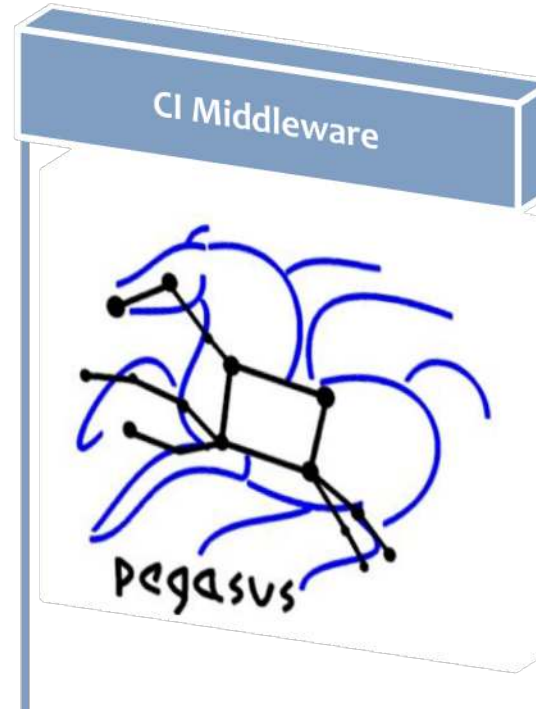


First GW detection: Pegasus automated ~ 21K workflows with ~ 107M tasks

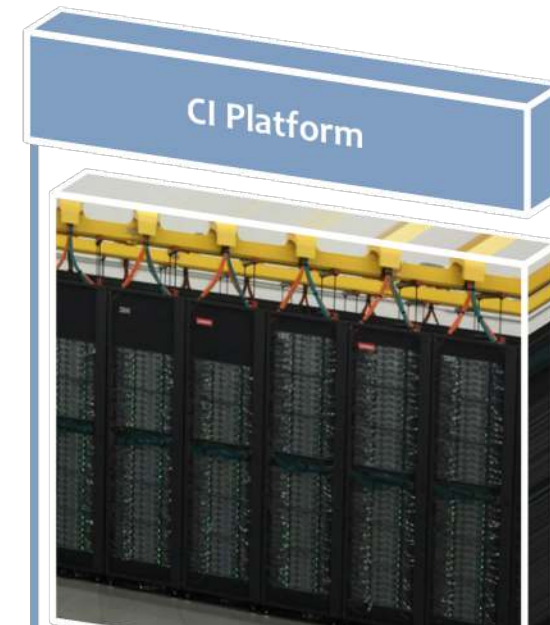
*Science workflow:
measure the statistical significance
of data needed for discovery*



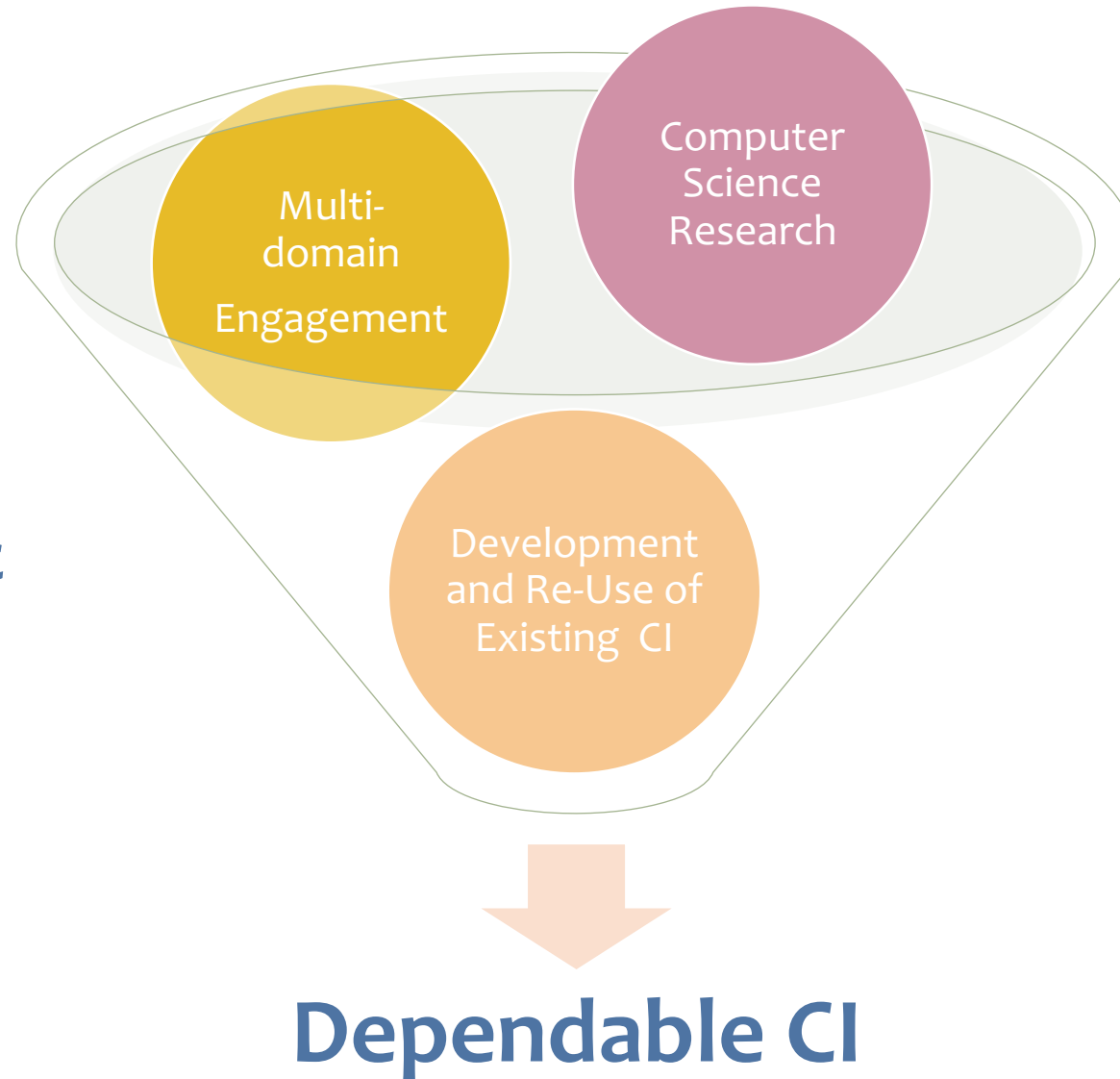
*Automated by Pegasus
execution of tasks and data
access*



*Distributed Power
LIGO, Open Science Grid,
XSEDE, Blue Waters*



What does it take to build and sustain Cyberinfrastructure?



**The Pegasus lesson-
One needs a holistic
approach to build
dependable CI!**



Takes time to build a team and expertise



Back Row: Tu Mai Anh Do, Mats Rynge, Karan Vahi, George Papadimitriou

Front Row: Rosa Filgueira, Ewa Deelman, Rajiv Mayani

Missing: Rafael Ferreira da Silva, Ashwin Venkatesha



Takes Contributions from Many People

GRAs



PostDocs



Master Students and Visitors



Developers



Currently at Amazon, Google, NetApps, SpaceX, Samsung, startups


And others

Takes Collaboration with Many CS and Domain Scientists



How did Pegasus Start?


Extend the concept of view materialization in DBs to distributed environments



The Virtual Data Grid (VDG) Model

- Data suppliers publish data to the Grid
- Users request raw or derived data from Grid, without needing to know
 - Where data is located
 - Whether data is stored or computed

NSF ITR: GriPhyN Project: Ian Foster (PI), Paul Avery, Carl Kesselman, Miron Livny, (co-PIs)



Virtual Data Scenario

- (LIGO) "Conduct a pulsar search on the data collected from Oct 16 2000 to Jan 1 2001"
- For each requested data value, need to
 - Understand the request
 - Determine if it is instantiated; if so, where; if not, how to compute it
 - Plan data movements and computations required to obtain all results
 - Execute this plan

How do you translate the Computer Science idea to the needs of science?

Circa. 2001



Challenge: How Translate a Science Request to an Actionable Plan?

Welcome to the LIGO-GriPhyN Prototype Demo.

LIGO Laser Interferometer Gravitational-Wave Observatory

GriPhyN Data Intensive Science

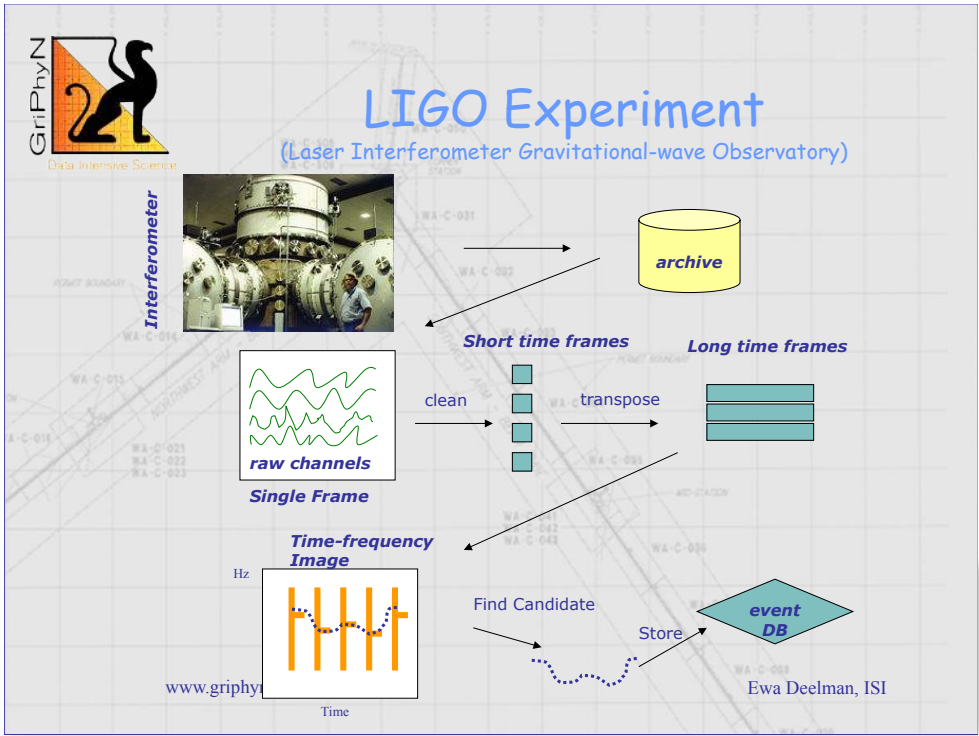
Please Enter Input Parameters below.

| | |
|---|--|
| Channel Name | H2:LSC-AS_Q |
| Start Time in GPS <small>(m692740114)</small> | 65800000 |
| End Time in GPS <small>(m693042393)</small> | 65800010 |
| Select Request Manager | <input checked="" type="radio"/> Execute this request <input type="radio"/> Echo this request |
| Select Output data Location (select server, type filename) | isi.edu (Los Angeles) file.xml |
| SUBMIT | Reset |

Completion Date November 2001

www.griphyn.org Ewa Deelman, ISI

Explore AI
planning
techniques



Lost in translation: high-level abstraction for this science domain
Found: new research direction: management of workflows in distributed environments

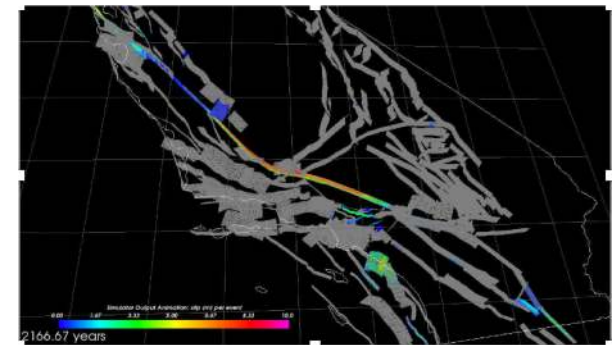


Challenges of Workflow Management

- Working with LIGO and other applications (astronomy, earthquake science), found common challenges:
 - Need to describe complex workflows in a simple way
 - Need to access distributed, heterogeneous data and resources
 - Need to deal with resources/software that change over time
- Our focus:
 - Separation between workflow description and workflow execution
 - Workflow planning and scheduling (scalability, performance)
 - Task execution (monitoring, fault tolerance, debugging)



Sky mosaic, IPAC, Caltech



Earthquake simulation, SCEC, USC

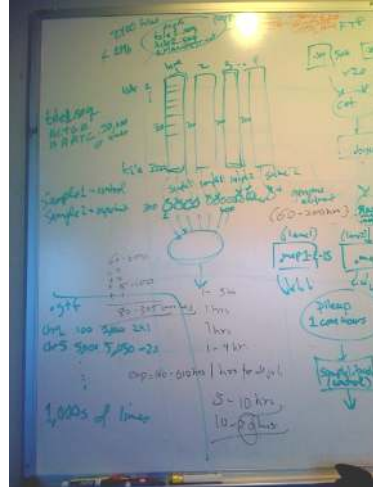


Benefits of Scientific Workflows (from the point of view of an application scientist)

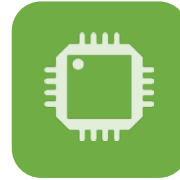
- Conducts a series of computational tasks
- Chaining (outputs become inputs) replaces manual hand-offs
- Ease of use: gives non-developers access to sophisticated codes
- Provides framework to host or assemble community set of applications, can be multi-disciplinary
- Framework to define common formats or standards when useful



Typical local computational environment



Work Definition



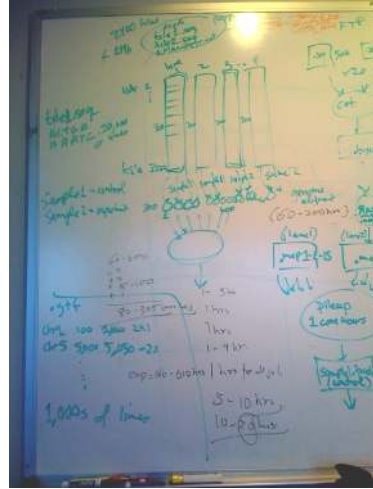
Local Resource



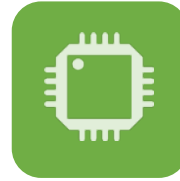
Local
Data
Storage



Typical local computational environment



Work Definition

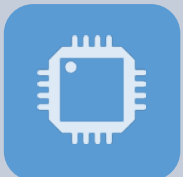


Local
Resource

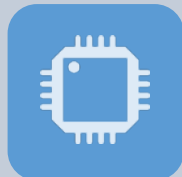


Local
Data
Storage

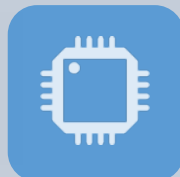
Blue
Waters



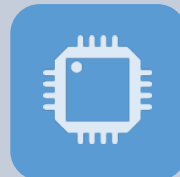
Campus
Cluster



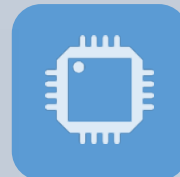
XSEDE



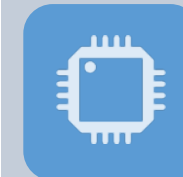
DOE
Facilities



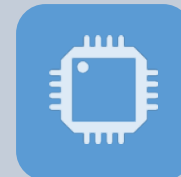
OSG



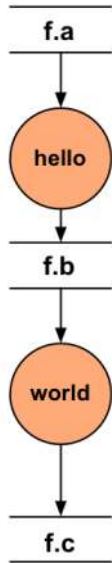
Chameleon



Amazon Cloud



To run Hello World on TACC's Wrangler



1. Login to TACC

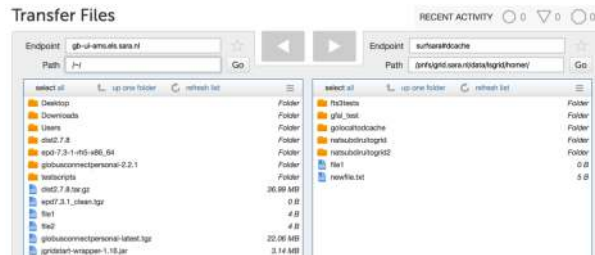
```
localhost$ ssh -l deelman wrangler.tacc.utexas.edu
login1.wrangler$ emacs myjob.sub
```

2. Write submit script

```
#!/bin/bash
#SBATCH -J myjob
#SBATCH -o myjob.o%j
#SBATCH -e myjob.e%j
#SBATCH -p normal
#SBATCH -N 1
#SBATCH -n 1
#SBATCH -t 01:30:00
#SBATCH --mail-
user=deelman@gmail.com
#SBATCH --mail-type=all
#SBATCH -A myproject

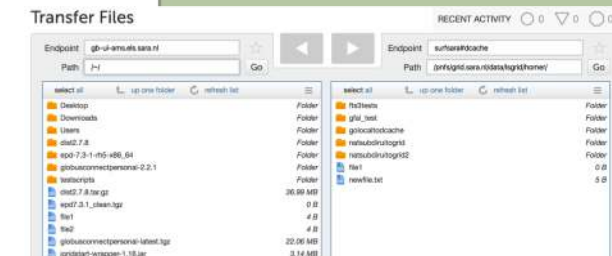
mkdir $WORK/helloworld
cd $WORK/helloworld
cp $WORK/data/inputs/* .
~/hello
~/world
cp * $WORK/data/outputs/
~/my_output_files/
```

3. Find and bring in your input data



4. Submit script for execution

```
login1.wrangler$ squeue myjob.sub
```

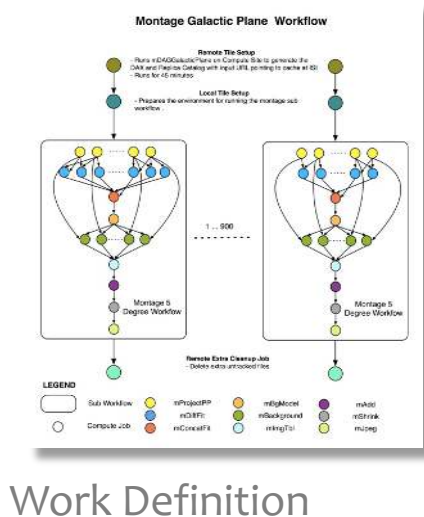


5. Stage out data for further analysis

What if Wrangler goes down/gets decommissioned? What if the job crashed? What about running on multiple platforms?



Our Approach: Submit locally, Compute globally



Local Resource



Workflow Management System



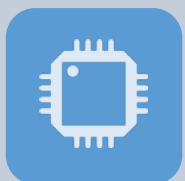
Local Data Storage

 High Throughput Computing

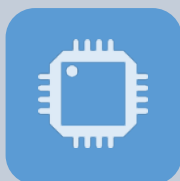
work

data

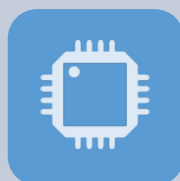
Blue Waters



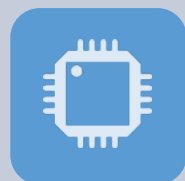
Campus Cluster



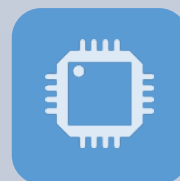
XSEDE



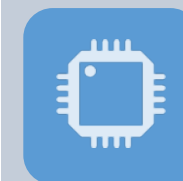
DOE Facilities



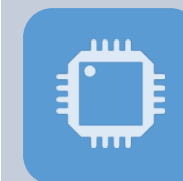
OSG



Chameleon

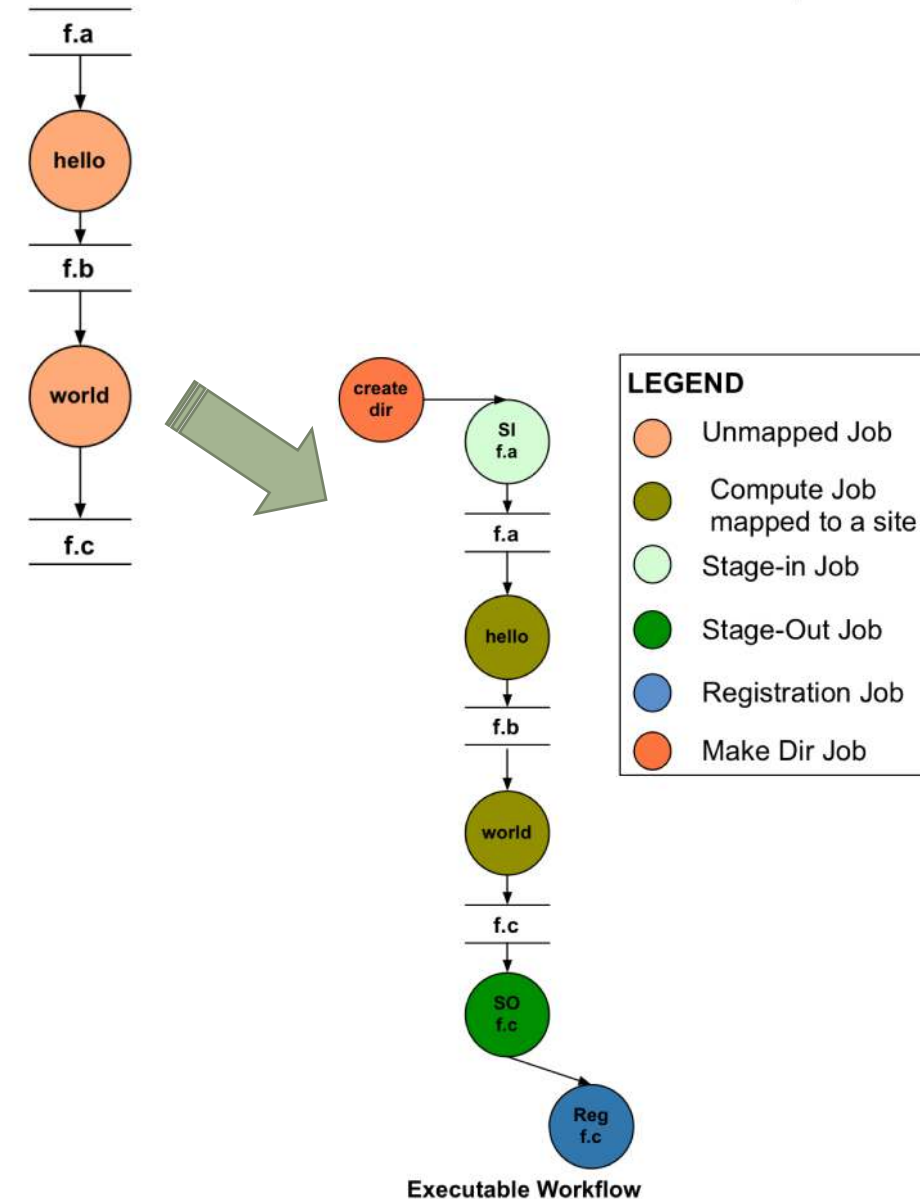


Amazon Cloud



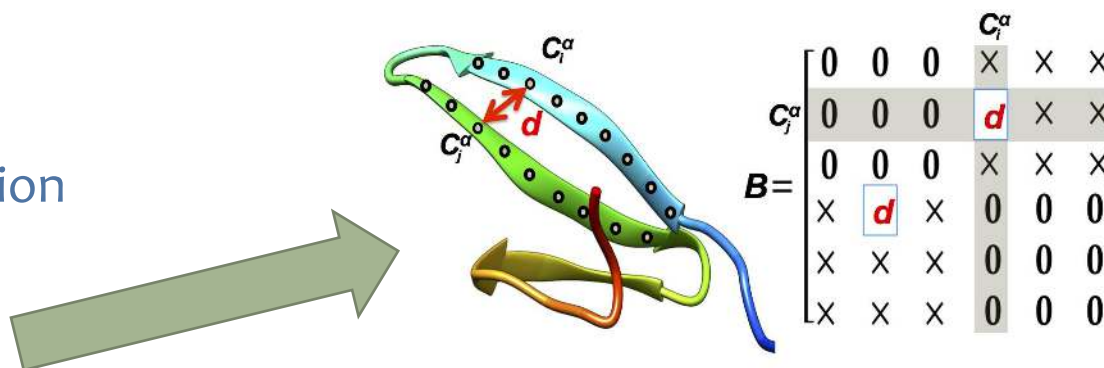
Pegasus Today

- Scientists describe their computational processes (workflows) at a logical level, without including details of the underlying CI
 - Operates at the level of files and individual codes
- Pegasus maps the abstract workflow to the available resources and infers the needed data transfers
- Pegasus generates an executional workflow, writes out submit files, and executes the workflow
- Underpins other user facing portals: NanoHub (Purdue)
- Provides workflow management for workflow composition tools: Wings (USC)



CS Principles Help in Cyberinfrastructure Development

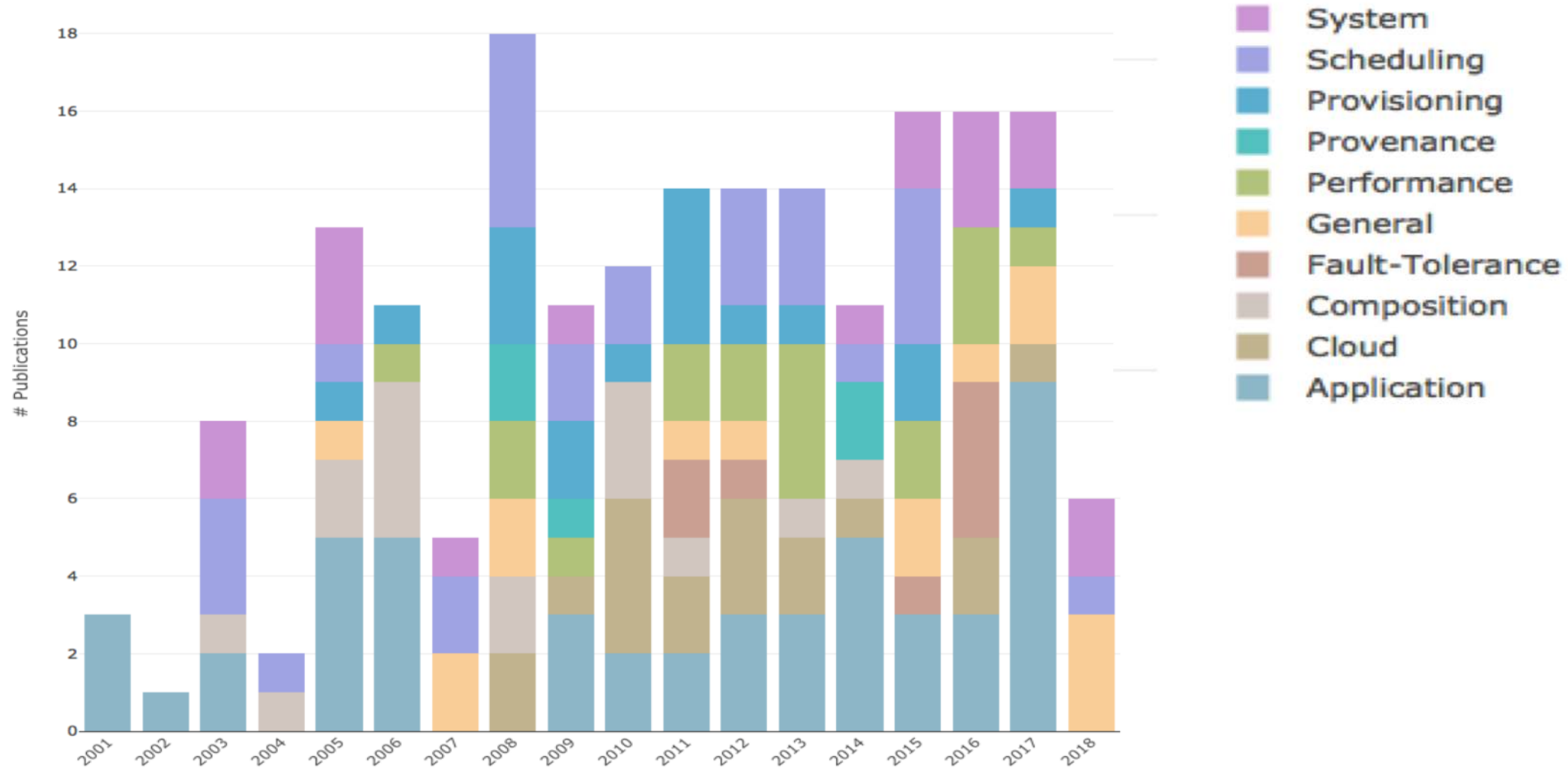
- Structure workflows as **directed acyclic graphs (DAGs)**
 - Re-use of graph traversal algorithms, node clustering, pruning, other complex graph transformation
- Use hierarchical structures in DAGs
 - To achieve scalability, recursion, dynamic behavior
- Develop new algorithms:
 - Task clustering
 - Data placement
 - Data re-use
 - Resource usage estimation
 - Resource provisioning
 - ***In situ* workflows**



New Direction:
In-memory coupling
of simulation and
analytics
Collaboration with U
Delaware, Cornell,
UTEP



Publications are important for dissemination, education, workforce development, career path, and funding



Publications over the years

Leveraging Proven Solutions Key to Innovation

- **Leveraged HTCondor's**

- Job submission to heterogeneous, distributed resources
- Managing job dependencies expressed as DAGs
- Job retries and error recovery

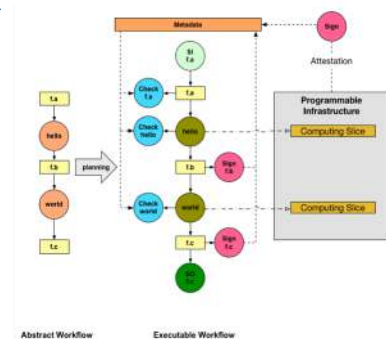


- **Allowed us to focus on other aspects of automation:**

- Workflow planning, and re-planning in case of failures
- Automated data management
- Specialized workflow execution engines for HPC systems
- APIs for workflow composition: Python, R, Java, Perl, Jupyter Notebook
- User-friendly monitoring and debugging tools
- Provenance tracking
- **Data integrity**

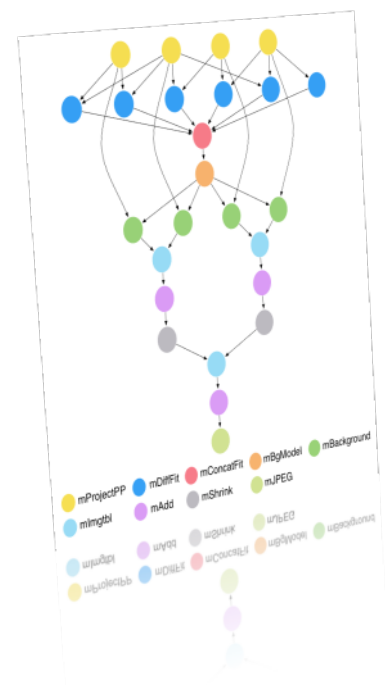


Indiana University
RENCI

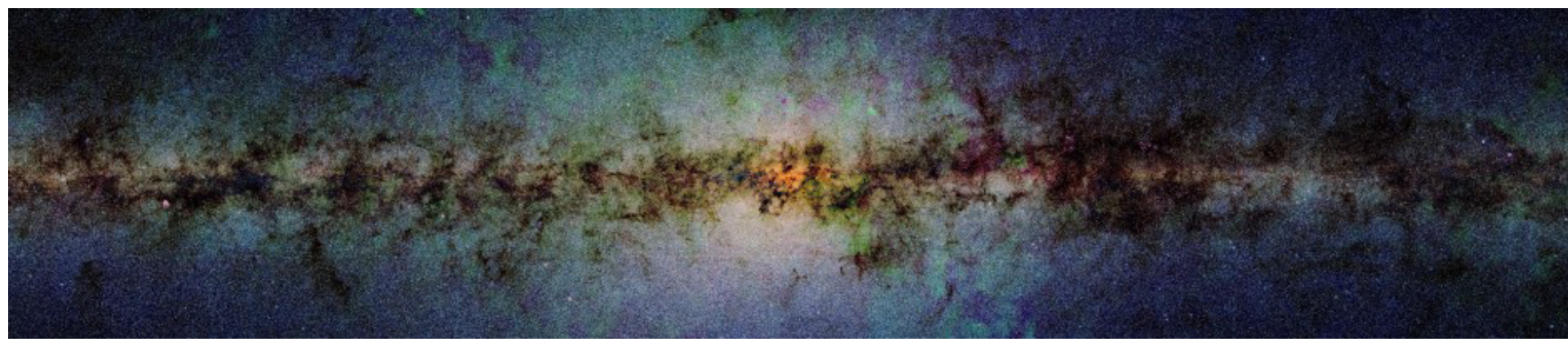


Using Real Applications Provides Realistic Testing and Evaluation

- Montage: Important application for CS and CI
- Open source, open data, scalable, robust
- Helps advance CS and test CI: workflow scheduling, resource provisioning, provenance tracking
- One of the workflows used in Pegasus' nightly build and test

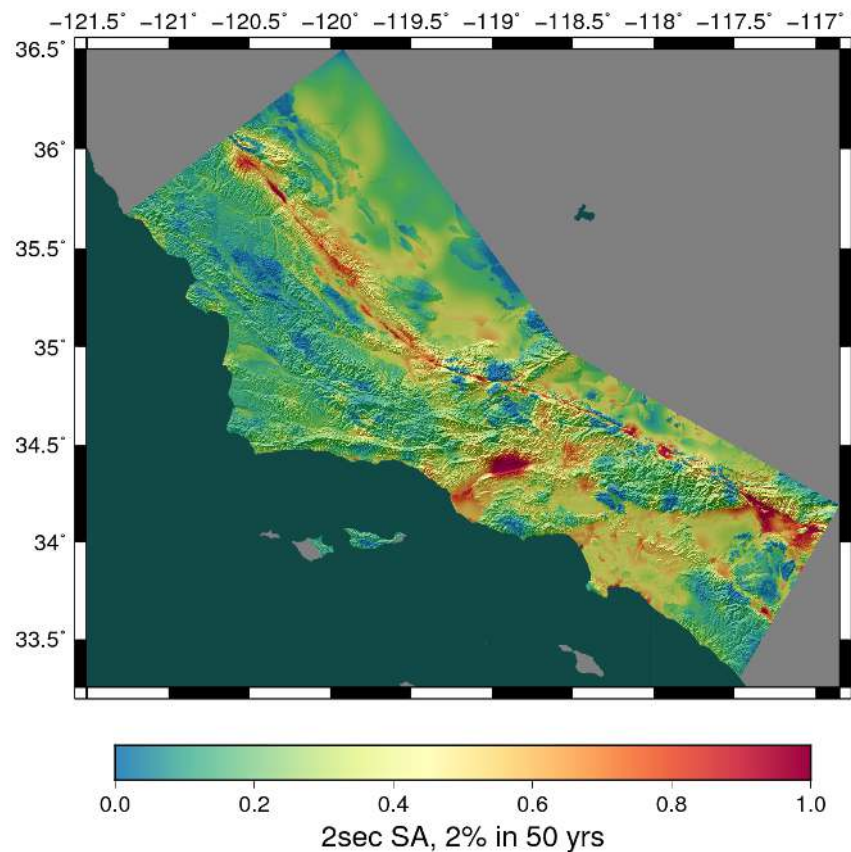


Montage, an important astronomy application, collaboration with Caltech since 2002



Need applications that push the boundaries of what you can do

SCEC's CyberShake: What will peak earthquake shaking be over the next 50 years?



Useful information for:
Building engineers
Disaster planners
Insurance agencies

2017: 21.6 million core hours, 777TB of data

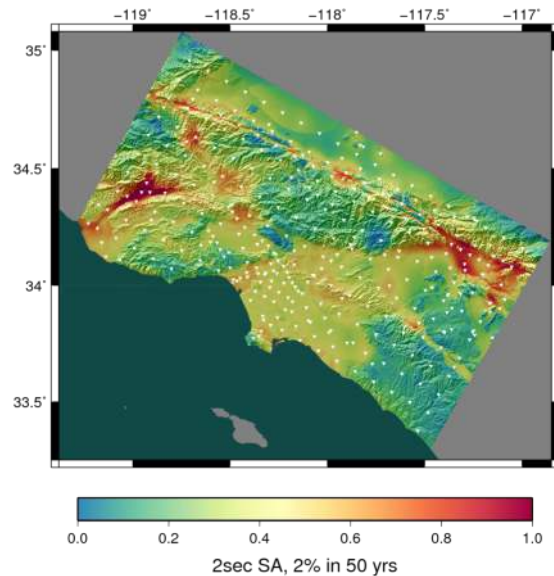
On ORNL's Titan and NCSA's Blue Waters



Since 2007: CyberShake ran on 9 different HPC systems 100 million core-hours (11,416 years)

Pegasus Optimizations:

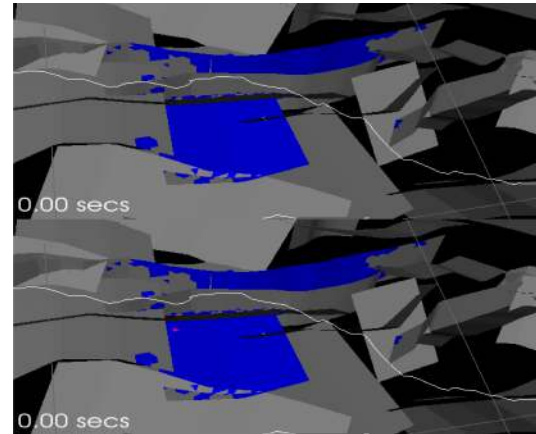
- Task clustering
- MPI-based workflow engine



2010: World's first physics-based probabilistic seismic hazard map,

Application Optimizations:

- Workflow restructuring
- MPI/code tuning
- Porting to GPUs



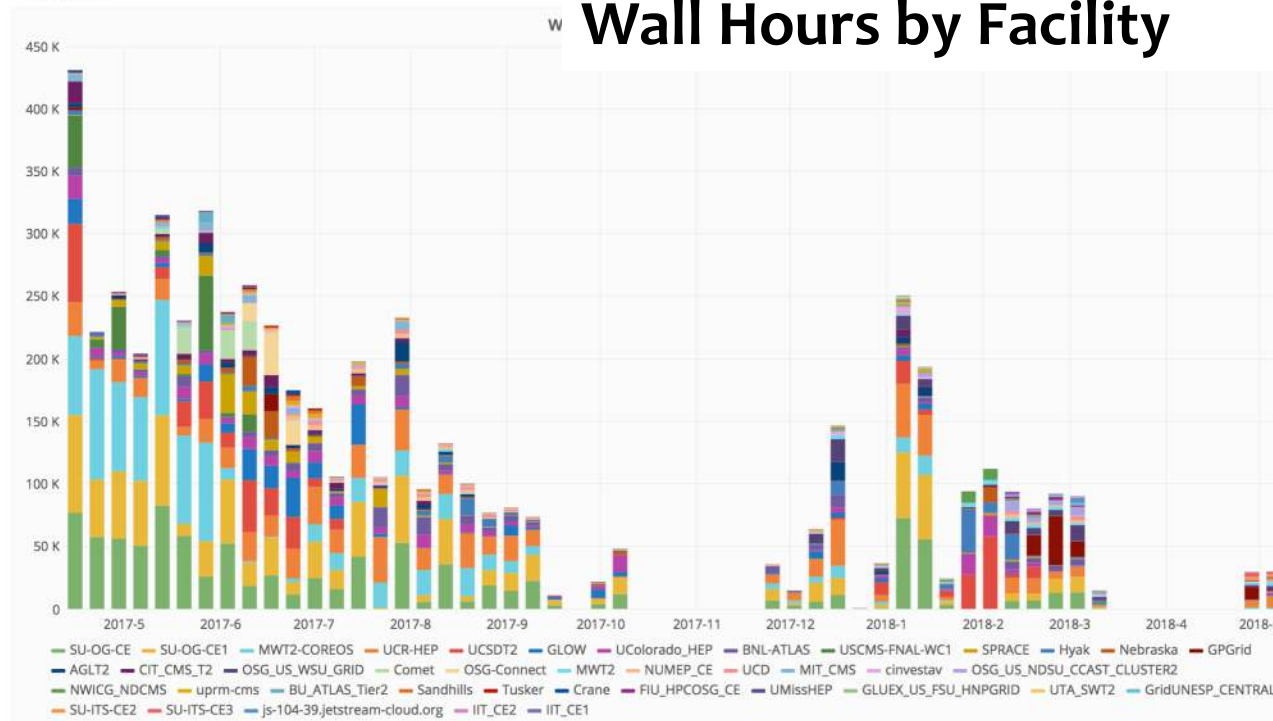
2018: Incorporating earthquake simulator with a 1 million-year catalog of California seismicity

USC HPCC

NCSA
MercuryTACC
RangerNICS
KrakenTACC
StampedeNCSA
AbeORNL
TitanNCSA
Blue
WatersSDSC
DataStar

Arming Individual Scientists with Pegasus on OSG

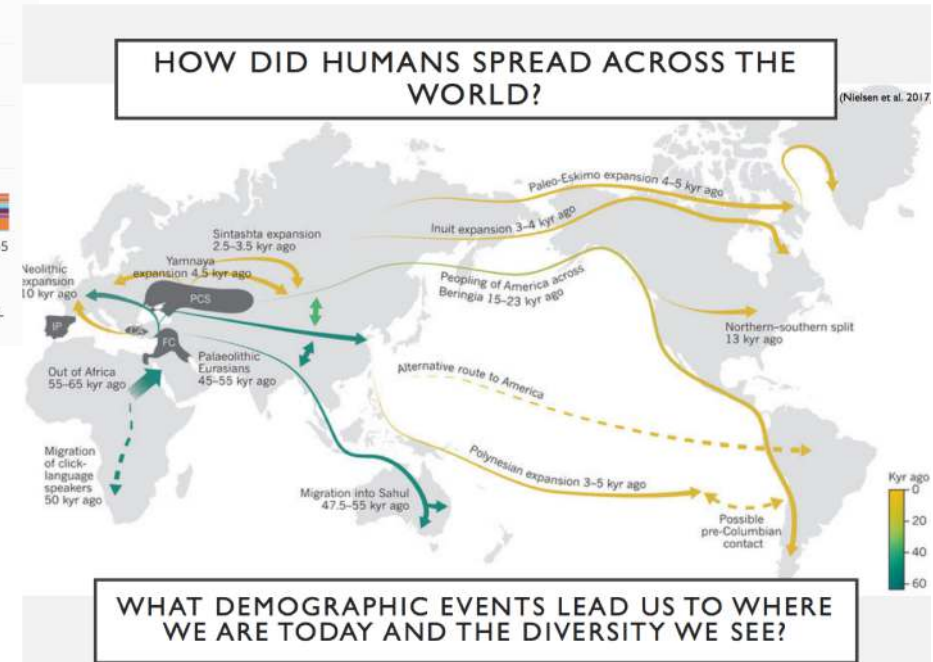
By Facility



Wall Hours by Facility

342 workflows
12 million jobs
40 execution sites
~ 7.3 Million Wall Hours

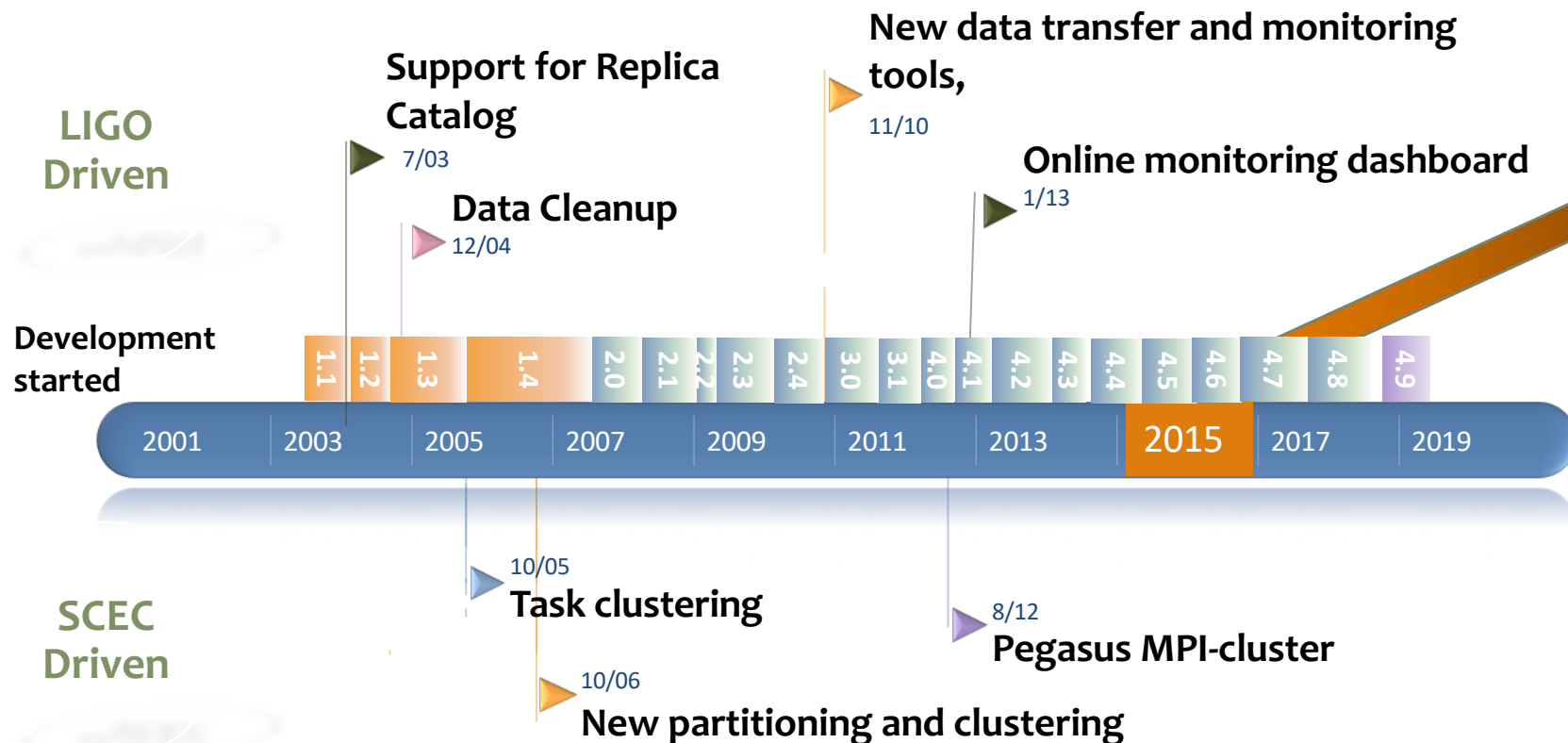
Ariella Gladstein, Ph.D. Student
University of Arizona



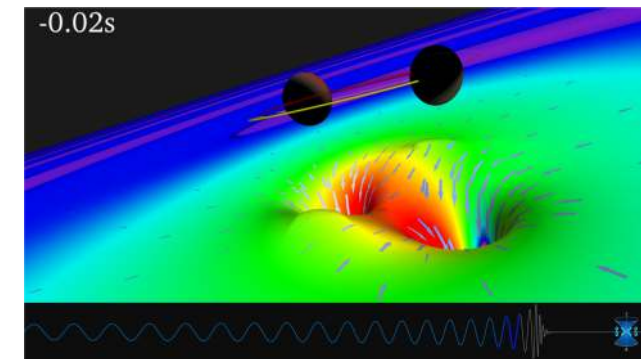
WHAT DEMOGRAPHIC EVENTS LEAD US TO WHERE WE ARE TODAY AND THE DIVERSITY WE SEE?



Cross-pollination between domains is highly beneficial



LIGO's first GW detection



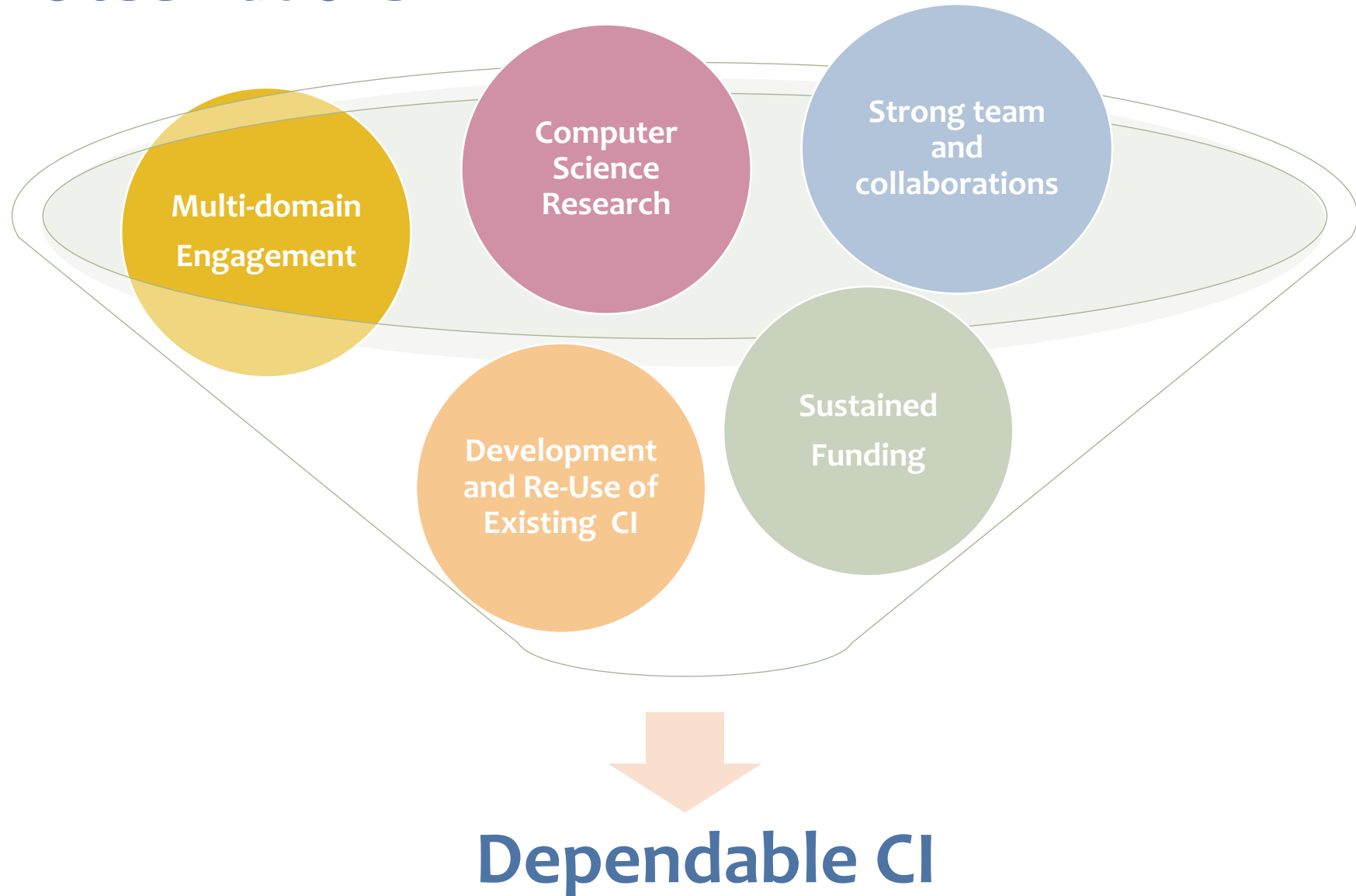
0.2 Second before a pair of massive black holes collide

Benefits the applications
Benefits the software

But, can make the
software more complex

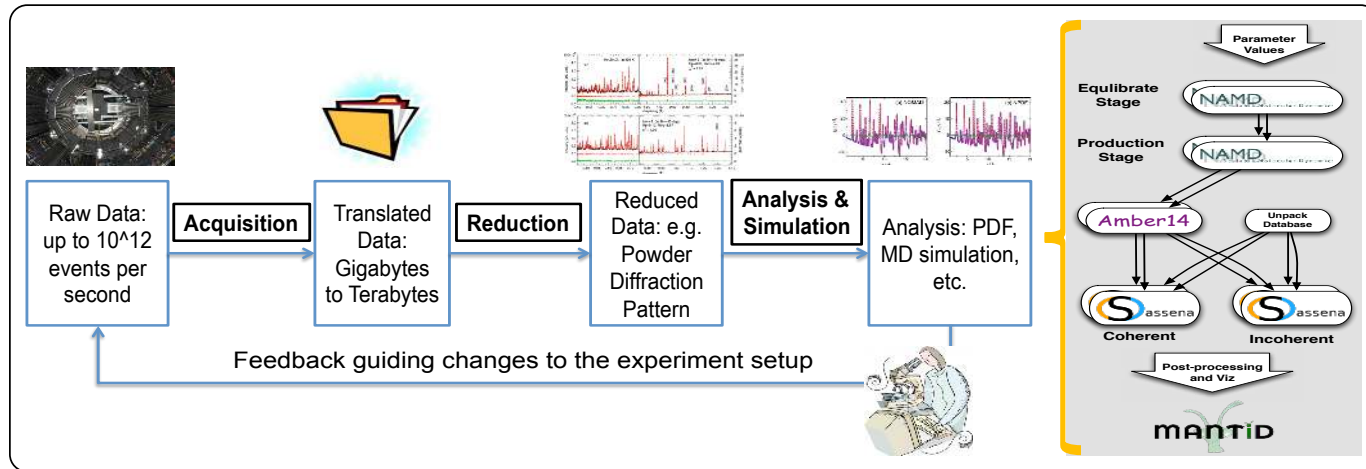


Summary of Observations



Looking ahead: Application Trends

Spallation Neutron Source



- More complex
- Faster time to solution: instrument steering
- More individual researchers in need of significant CI
 - Need intuitive workflow composition, better monitoring, error handling, assisted debugging

Planned CyberShake for Northern California:

- 869 geographic sites
- 16,000 workflow jobs
- 70 million core-hours on Blue Waters and Titan
- 800 TB of data

Outreach: How do you reach scientists that don't know you are out-there?

Many scientists are going through the same pain

Leverage/enhance existing engagement: NSF's Campus efforts? OSG/XSEDE outreach?

Education and outreach at instruments and experimental facilities?



Looking ahead: Growing Demand for Automation

HPC Systems

- Complex
- Heterogeneous
- Specialized data storage
- Increasingly faulty

Distributed Systems

- Software Defined capabilities
- Specialized data storage

Clouds

- New platform for science
- Very heterogeneous
- Can be costly

Resource Management is Key

Constraints: time to solution, budget

Faulty environment: failure detection and attribution

Heterogeneous storage: memory, burst buffers, file systems, data xfer nodes

Need to keep track of big data technologies and machine learning solutions that are being developed at a rapid pace by industry

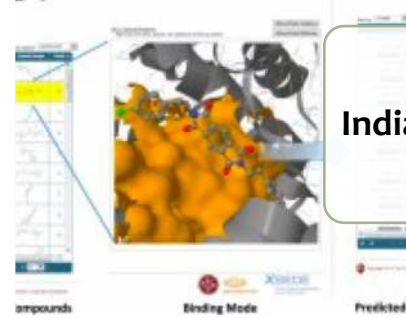


Cancer genetics



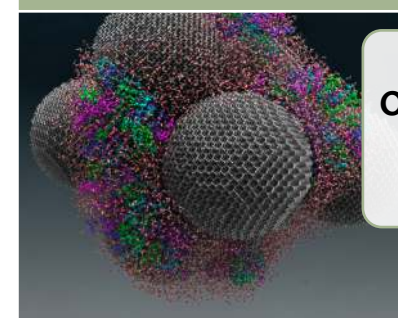
Clemson

Structural Protein-Ligand Interactome



Indiana U.

Molecular Dynamics

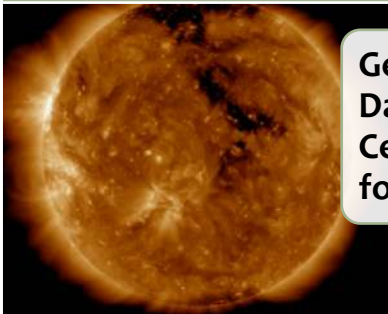


ORNL

Thank you to
the team,
collaborators
and funders

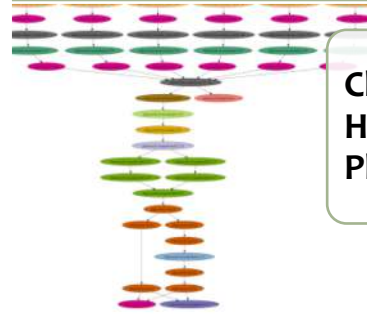
Example
Pegasus-
enabled
Applications

Helioseismology



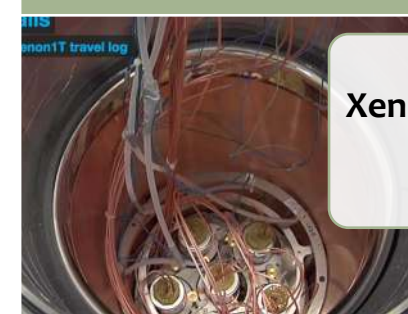
German
Data
Center
for SDO

Next Generation Sequencing



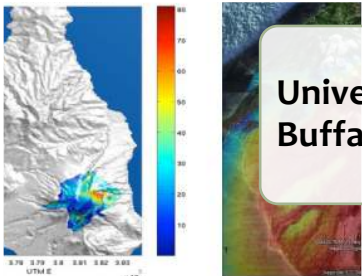
Children's
Hospital of
Philadelphia

Dark Matter Detection



Xenon1T

Volcano Mass Flow



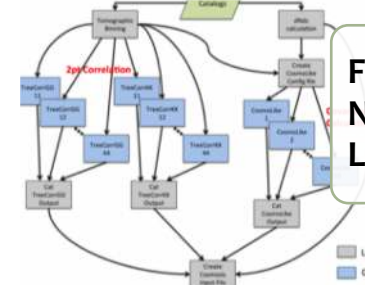
University of
Buffalo

Soybean Studies



University
of Missouri

Weak Gravitational Lensing



Fermi
National
Lab

We look
forward to
future
collaborations