Workflow Tools

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Information Sciences Institute

Overview

- What are scientific workflows and why use them?
- Example workflow applications
- Overview of available workflow systems
- Introduction to Pegasus WMS







Scientific Workflows

- Formal way to capture multi-step computations
- Relatively coarse grained
- Capture the steps and their parameters
- Define the input/output data of each step
- Describe dependencies between steps





Workflows can be simple!







Why Scientific Workflows?

- Automate complex, multi-stage processing pipelines
- Enable parallel, distributed computations
- Use existing code, no rewrites
- Simple to construct and modify
- Reusable, aid reproducibility
- Can be shared with others
- Record how data was produced (provenance)
- Handle failures with to provide reliability
- Keep track of data and files





Science-grade Mosaic of the Sky

Science-grade Mosaic of the Sky



Montage Workflow

montage.ipac.caltech.edu

Si	ize of mosai in degrees square	c Number of input data files	Number of tasks	Number of intermediate files	Total data footprint	Cummulative wall time
	1 3	84	387	850	1.9 GB	21 mins
	2	300	1442	3176	6.8 GB	54 mins
	4	685	3738	8258	18 GB	3 hours, 18 mins
	6	1461	7462	16458	37 GB	7 hours, 7 mins
	8	2565	12757	28113	64 GB	11 hours, 44 mins

Bag of Tasks: Periodogram Workflow

- Kepler continuously monitors the brightness of over 175,000 stars
 - Search for periodic dips in signals as Earth-like planets transit in front of host star.
- For each star, Kepler data is used to create a "light curve"
- Need to perform a bulk analysis of all the data to search for these periodic signals

2012 Run at SDSC

- 1.1M tasks, 180 jobs
- 1.1M input, 12M output files
- ~101,000 CPU hours
- 16 TB output data



Kepler 6-b transit

exoplanetarchive.ipac.caltech.edu

Workflows with MPI Codes: Neutron Scattering



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- Spallation Neutron Source at ORNL
 - Parameter sweeps of MD and neutron scattering simulations
 - Fit simulation to experimental data
 - e.g. temperature, charge, force
- Nanodiamond Workflow
 - Feb 2015 on Hopper using GRAM and GridFTP
 - 19 parameter values for nonbonded interactions between ND and H2O
 - 800 core NAMD jobs x 22 hrs
 - 400 core Sassena jobs x 3 hrs
 - ~380,000 CPU hours
 - ~1/2 TB output



Large-Scale Workflows: CyberShake PSHA



2014: 286 Sites, 4 models

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Each site = one workflow

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- Each workflow has 420,000 tasks in 21 jobs using task clustering w/ PMC
- BlueWaters@NCSA, Stampede@TACC

- Builders ask seismologists: "What will the peak ground motion be at my new building in the next 50 years?"
- Seismologists answer this question using Probabilistic Seismic Hazard Analysis (PSHA)



Workflow Management Systems

- Automate execution of workflows
- Workflow language
 - Used to describe the workflow
 - Visual with GUI or text-based
 - Frequently based on DAGs, but some provide loops and branches or more exotic semantics
- Workflow engine
 - Manages the scheduling, submission, and monitoring of tasks
 - Orchestrates the movement of data
 - Interfaces with diverse cyberinfrastructure (grids, clusters, clouds)
- There are lots of workflow management systems
 - Some are abandoned research projects





Swift (swift-lang.org)



- Developed at the University of Chicago
- Workflow defined via parallel scripting language

```
//Create new type
type messagefile;
//Create app definition, returns messagefile
app (messagefile t) greeting() {
    //Print and pipe stdout to t
    echo "Hello, world!" stdout=filename(t);
}
//Create a new messagefile, linked to hello.txt
messagefile outfile <"hello.txt">
//Run greeting() and store results
outfile = greeting();
```

- Supports workflows with many tasks and large data
- Interfaces with many different cluster, grid and cloud infrastructures





Kepler (kepler-project.org)



- Developed by a diverse group of collaborators
- GUI-based
 - Composition and execution
 - View outputs
- Many different models of computation
 - Actor model with different execution semantics
- Interfaces with grids, clusters, and web services
- Component repository for sharing and lots of built-in components







Taverna (www.taverna.org.uk)



- Developed by a collaboration of UK universities
- GUI workflow composition
 - DAGs, loops, data parallel, merges
- Web services and local scripts/commands (mostly)
- Particularly good for bioinformatics
- Integrates with myExperiment for sharing workflows
- Leverages service catalogs for easy workflow composition

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WS-PGRADE/gUSE (guse.hu)

- Developed at the Hungarian Academy of Sciences
- GUI interface for workflow composition
- Supports template DAGs for parameter sweep, WoW

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- Integrated web portal/gateway
- Interfaces with many different infrastructures
- Extensive documentation



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Other Workflow Systems

- VisTrails (vistrails.org)
 - Used for visualization pipelines with VTK
- Galaxy (galaxyproject.org)
 - Oriented toward biomedical research
 - Interfaces with many web services
 - Web-based GUI interface
- UNICORE Workflow System (unicore.eu)
 - GUI for workflow composition, or XML
 - Branches, loops, parallel loops
- Makeflow (ccl.cse.nd.edu/software)
 - Simple, make-like workflow language
 - Targets many different grid, cluster systems

















Pegasus Workflow Management System

Under development since 2001

- Pegasus
- A collaboration between USC/ISI and the Condor Team at UW Madison
 - USC/ISI develops Pegasus
 - UW Madison develops DAGMan and Condor
- Actively used in a wide variety of domains
 - Earth science, physics, astronomy, bioinformatics, climate modeling, neutron science, and many others
 - About 600 workflows a day





Why Pegasus?

- Maps abstract workflows to diverse computing infrastructures
 - Desktop, Condor Pool, HPC Cluster, Grid, Cloud
- Supports large-scale, data-intensive workflows
 - O(1M) tasks and O(TB) of data
- Automatically plans and executes data transfers
- Manages failures to provide reliability
 - Retries and checkpointing
- Provides workflow monitoring and debugging tools to allow users to debug large workflows
- Technical support
 - full-time staff, mailing lists, public repository and bug tracker, regular releases, decent documentation





Pegasus Workflows

- Expressed as a DAG: nodes=tasks, edges=dependencies
- Tasks are command-line programs, executed as batch jobs
- Dependencies are usually data dependencies
- Data is exchanged via files



Pegasus WMS Environment





Pegasus WMS Data Flow



Workflow Planning (Mapping)

- Pegasus converts abstract workflow descriptions into executable workflows (similar to compiler)
 - Facilitates portability
 - Separates data management from workflow composition
 - Enables workflow-level optimizations
 - Others...
- Planning process:
 - Choose a site for each job (site selection)
 - Add resource-specific information
 - Choose input files (replica selection)
 - Plan data movements and add data management jobs
 - Perform optimizations
 - Add setup and cleanup jobs
 - Generate executable workflow artifacts





Abstract to Executable Workflow Mapping





Data Management



- Most of the tasks in scientific workflow applications require **POSIX** file semantics
 - Each task in the ____ workflow opens one or more input files
 - Read or write a portion _ of it and then close the file.
- Data Staging Site can be the shared filesystem on the compute cluster!

Task Flow

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Data Flow

- Input Data Site, Compute Site and Output Data Sites can be co-located
 - Example: Input data is already present on the compute site.



Data Staging Configurations

Shared File System (typical of most HPC sites)

- Worker nodes and the head node have a shared filesystem, usually a parallel filesystem with high-performance I/O
- Can leverage symlinking against prestaged datasets
- Staging site is the compute site



Non-shared File System (typical of OSG and EC2)

- Worker nodes don't share a file system
- Uses a staging site separate from the compute site such as Amazon S3
- Data is pulled from / pushed to the staging site
- Also known as "PegasusLite"





Data Staging Configurations

Condor I/O (Typical of Condor Pools like OSG sites)

- Worker nodes don't share a file system
- Data is pulled from / pushed to the submit host via Condor file transfers
- Staging site is the submit host

Using Pegasus allows you to move from one deployment to another without changing the workflow description

Many Data Protocols Supported:

- SCP
- HTTP

- GridFTP
- Amazon S3 iRODS
- FTP

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- symlink
- **FDT**

Jobs

- Google Storage
- **StashCache**





Workflow Monitoring and Reporting

Data collection

- Data extracted from log files and stored in a relational database
- DB contains workflow structure, status information, runtimes, host info, task stdout/stderr

Reporting tools

- Status of the workflow
 - pegasus-status path/to/submit/directory
- Detailed runtime statistics
 - pegasus-statistics -s all path/to/submit/directory

Туре	Succeeded	Failed	Incomplete	Total	Retries	Total+Retries				
Tasks	135002	0	0	135002	0	135002				
Jobs	4529	0	0	4529	0	4529				
Sub-Workflows	2	0	0	2	0	2				

Workflow wall time: 13 hrs, 2 mins, (46973 secs)Workflow cumulative job wall time: 384 days, 5 hrs, (33195705 secs)Cumulative job walltime as seen from submit side: 384 days, 18 hrs, (33243709 secs)



Pegasus Dashboard

- Web-based workflow monitoring GUI
 - Data comes from monitoring database
 - Supports monitoring, troubleshooting, and reporting



Failure Management

- Pegasus detects job failures
 - non-zero exit code
 - output does not contain a specified "success message"
 - output does contain a specified "failure message"
 - it exceeds a specified time limit
 - it fails to produce expected output files
- Job Retries
 - Helps with transient failures
 - Each job has a set number of retries per run
- Rescue DAGs
 - DAGMan writes a checkpoint file so workflow can be restarted
 - Can recover from almost any failure with minimal loss
- Checkpoint files
 - Job generates checkpoint files
 - Staging of checkpoint files is automatic on restarts





Workflow Debugging

- Problem: You have 1M tasks, and one of them fails
- pegasus-analyzer: Provides summary of workflow execution
- Outputs
 - A brief summary section
 - showing how many jobs have succeeded
 - and how many have failed
 - For each failed job:
 - showing its last known state
 - exitcode
 - working directory
 - the location of its submit, output, and error files
 - any stdout and stderr from the job





Task Clustering

- Cluster small running jobs together to achieve better performance
- Why?
 - Each job has scheduling overhead need to make this overhead worthwhile
 - Ideally users should run a jobs that take at least 10/30/60/? minutes
 - Clustered tasks can reuse common input data less data transfers



Pegasus-MPI-Cluster

- A master/worker task scheduler for running fine-grained workflows and ensembles on HPC systems
- Runs as an MPI job \rightarrow Works on most HPC systems
- Allows sub-graphs of a Pegasus workflow to be submitted as monolithic jobs to remote resources
- Can be used on a sub-graph, or the entire workflow



PMC Features

- Fault Tolerance
 - Retries at the task level (master resends task to another worker)
 - Retries at the workflow level (using a transaction log to record progress)
- Resource-aware scheduling
 - Many HPC machines have low memory/core
 - PMC can allocate memory and cores to a task, and force other slots on the same node to be idle
- I/O Forwarding
 - Small tasks == small I/O == poor performance
 - PMC reads data off of pipes from worker and forwards it using MPI messages to a central I/O process, which collects the data and writes it to disk
 - Writes are not interleaved, no locking required for synchronization





Resource Provisioning with Pilot Jobs

- Key idea: Use HPC scheduler to run application scheduler
- Parallel pilot jobs
- Amortize queue delays over many application jobs
- Apply applicationspecific policy







Data Cleanup



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Workflow Reduction (Data Reuse, Restarts)





Data reuse happens automatically when output files are found in the replica catalog



Large-scale, Hierarchical Workflows







Other Features

- Job and Transfer Throttling
 - Prevents too many jobs/transfers from overloading system
- Notifications
 - System calls a script when certain events occur: send email, text, etc.
- Executable and Worker Package Staging
 - Enables dynamic deployment of code on remote sites
 - Planner matches the executable in the TC to the site in the SC
- Kickstart Job Wrapper
 - Records detailed information about job execution (execution host, environment, memory usage, I/O, files accessed, CPU time, etc.)
- Shell planner mode
 - Generate a shell script of your workflow





Final Thoughts

- Probably using a workflow already
 - Replaces scripts, manual hand-offs and polling to monitor
- Automation is vital
 - Eliminate babysitting your jobs: your time is valuable!
 - Able to recover from failures without losing work
- Put ALL processing steps in the workflow
 - Include validation, visualization, data publishing, notifications
- Does add additional software layers and complexity
 - Some development time is required
- Choose workflow system carefully
 - Consider required features, target environment, maturity, support
- We want to help you!





Questions?





Some Computational Science Challenges

- Integrate several programs into one pipeline
- Run an ensemble of simulations
- Repeat processing steps on new data or parameters
- Reproduce previous results, or similar results
- Share analysis steps with other researchers
- Recreate the history of data products
- Run code on hundreds or thousands of inputs
- Execute analyses in parallel on distributed resources
- Reliably execute pipelines on unreliable infrastructure

Scientific workflows can help with these problems





Workflow Management System Functionality

- Job execution
 - Interfaces with middleware and batch systems to submit and monitor jobs
- Data and control dependencies between jobs
 - Tracks dependencies and makes sure jobs are executed in the right order
- Scheduling
 - Some jobs may be able to run in parallel with others
 - Ordering and placement can improve performance
- Data management
 - Transfers of input and output files to/from machine
- Provenance
 - Track when a job was run, where it was run, what data it produced, key parameters, metadata
- Reliability
 - Keeps track of what finished successfully, and what did not
- Resource provisioning
 - Allocating resources to run jobs





Example Workflow



Example DAX Generator in Python

```
# Create DAX object
dax = ADAG("test dax")
# Define first job
firstJob = Job(name="first job")
# Input and output files to first job
firstInputFile = File("input.txt")
firstOutputFile = File("tmp.txt")
# Args to first job (first job input=input.txt output=tmp.txt)
firstJob.addArgument("input=input.txt", "output=tmp.txt")
# Role of the files for the job
firstJob.uses(firstInputFile, link=Link.INPUT)
firstJob.uses(firstOutputFile, link=Link.OUTPUT)
# Add the job to the workflow
dax.addJob(firstJob)
```



```
for i in range(0, 5):
    # Create simulation job
    simulJob = Job(id="%s" % (i+1), name="simul job")
    # Define files
    simulInputFile = File("tmp.txt")
    simulOutputFile = File("output.%d.dat" % i)
    # Arguments to job
   # simulJob parameter=<i> input=tmp.txt output=output<i>.dat
    simulJob.addArgument("parameter=%d" % i, "input=tmp.txt",
       "output=%s" % simulOutputFile.getName())
    # Role of files
    simulJob.uses(simulInputFile, link=Link.INPUT)
    simulJob.uses(simulOutputFile, line=Link.OUTPUT)
    # Add job to dax
    dax.addJob(simulJob)
    # Dependency on firstJob
    dax.depends(parent=firstJob, child=simulJob)
# Write to file
fp = open("test.dax", "w")
dax.writeXML(fp)
fp.close()
```





Site Catalog

- Stores details about each target execution/storage site
 - Job submission endpoints (GRAM URL, etc.)
 - Paths to storage/scratch directories
 - Data transfer services (GridFTP servers, etc.)
 - Paths to credentials (X509 proxy, ssh key, etc.)
 - Site-level configuration (environment variables, etc.)
 - "local" site is special-refers to submit host

```
<!-- Example site catalog -->
<site catalog>
<site handle="example" arch="x86_64" os="LINUX">
<grid type="gt5" contact="example.isi.edu/jobmanager-fork" jobtype="auxillary"/>
<grid type="gt5" contact="example.isi.edu/jobmanager-pbs" jobtype="compute"/>
<directory type="shared-scratch" path="/scratch">
<file-server operation="all" url="gsiftp://example.isi.edu/scratch"/>
</directory>
<profile namespace="env" key="GLOBUS_LOCATION">/usr/local/globus</profile>
<profile namespace="pegasus" key="style">globus</profile>
<profile namespace="pegasus" key="X509_USER_PROXY">/tmp/x509_u40001</profile>
</site>
</site>
```





Transformation Catalog

- Maps transformations to executables on each site
 - Physical path or URL of executable and dependent data/ configuration files
 - Executable characteristics (OS, architecture, glibc, etc.)
 - Job-level configuration (e.g. environment variables, profiles)

```
# Example transformation catalog
tr example::date {
    profile env "TZ" "America/Los_Angeles"
    site example {
        pfn "/bin/date"
        os "linux"
        arch "x86_64"
        type "INSTALLED"
    }
}
```





Replica Catalog

- Maps logical files to physical files
 - LFN (name) to PFN (path or URL)
 - Mappings annotated with metadata (e.g. site/pool, size, etc.)
- Enables Pegasus to choose "best" replica (replica selection phase of planner)
- Where Pegasus registers workflow output locations
- Support file-based or DB-based RC (also callout)

```
# Example replica catalog
f.1 gsiftp://example.isi.edu/inputs/f.1 pool="example"
f.1 file:///inputs/f.1 pool="example"
f.2 file:///inputs/f.2 pool="example"
f.2 file:///inputs/f.2 pool="local"
```





Configuration Properties and Profiles

- Specify all the tuning knobs for Pegasus
- Unification of properties and profiles several years ago
- Often in a "pegasus.properties" file (or command-line)
- Some are global and apply to all sites and jobs
- Some (profiles) can also be specified in the TC, SC and DAX with different scopes
- Examples
 - pegasus.data.configuration = sharedfs
 - pegasus.style = condor
 - dagman.retry = 2
 - pegasus.exitcode.successmsg = "All data processed"



Data Management



- Compute site: compute jobs

Checkpoint Files

- A job can specify that it uses one or more checkpoint files
- Checkpoint files are both input files and output files
- Pegasus will stage-out these files in the case that job fails
 - Typically due to a timeout on long-running jobs
 - Jobs must periodically write checkpoint files (no signals are given)
- Pegasus will stage-in these files before retrying the job
 - They will appear in the working directory of the job



Workflow and Task Notifications

- Users want to be notified at certain points in the workflow or on certain events
- Support for adding notification to workflow and tasks
- Event based callouts
 - On Start, On End, On Failure, On Success
 - Examples contain email and jabber notification scripts
 - Can run any user provided scripts
 - Defined in the DAX





Pegasus clients for data management

- pegasus-transfer, pegasus-create-dir, pegasus-cleanup
- Support many different protocols
 - HTTP

- SCP

– Amazon S3

– SRM

- GridFTP
- IRODS cp
 - In -s
- Remote directory creation and removal
- Support client discovery, parallel transfers, retries, and many other things to improve transfer performance and reliability



Different Directories used by Pegasus

1. Submit Directory

- The directory where pegasus-plan generates the executable workflow i.e HTCondor DAGMan and job submit files.
- Specified by *--dir* option to pegasus-plan

2. Input Directory

- Mostly input file locations are catalogued in the Replica Catalog.
- However, if inputs are on the submit host, then you can pass –*input-dir* option to pegasus-plan

3. Scratch Directory

- Workflow specific directory created on the staging site by the create-dir job. This is where all the workflow inputs and outputs are gathered.
- The base directory specified in the site catalog entry in *sites.xml* file.

4. Output Directory

- The output directory where the outputs of the workflow appear.
- Specified in the output site entry in the *sites.xml* file.
- Can also be optionally specified by -output-dir option to pegasus-plan





Planning and Submitting workflows

pegasus-plan

- Interface to the Pegasus planner
- Specify input dir
- Specify compute site(s)
- Specify staging site(s)
- Specify output dir or output site

Pegasus-run

- Start and restart the workflow





Problems Workflows Solve

- Task executions
 - Workflow tools will retry and checkpoint if needed
- Data management
 - Stage-in and stage-out data
 - Ensure data is available for jobs automatically
- Task scheduling
 - Optimal execution on available resources
- Metadata
 - Automatically track runtime, environment, arguments, inputs
- Getting cores
 - Whether large parallel jobs or high throughput







Askalon (askalon.org)

- Developed at University of Innsbruck in Austria
- Create workflow description in AGWL (XML) or UML
 - if, for, parallelFor, DAGs
- Conversion: like planning, to bind to specific execution
- Submit jobs to Enactment Engine, which distributes jobs for execution at remote cluster, grid or cloud sites
- GUI for composition and monitoring







Example Hierarchical Workflow

<dax> element behaves like <job>

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- Arguments are for pegasus-plan (most are inherited)
- Planner is invoked when DAX job is ready to run

```
<?xml version="1.0" encoding="UTF-8"?>
<adag version="3.4" name="multi-level">
        <jobid="ID0000001"namespace="example"name="sleep">
                 <argument>5</argument>
        </job>
        <dax id="ID0000002" file="sub.dax">
                 <argument>--output-site local</argument>
        </dax>
        <jobid="ID0000003"namespace="example"name="sleep">
                 <argument>5</argument>
        </job>
        <child ref="ID000002">
                 <parent ref="ID0000001"/>
        </child>
        <child ref="ID000003">
                 <parent ref="ID0000002"/>
        </child>
</adaq>
```

Integration with HUBzero

A Brown, Jr. Network for Ea	arch, collaboration and education				
About NEES Too	Is & Resources Learning & Outreach Project Warehouse Simulation Sites Collaborate Explore NEEShub Supp				
You are here: The Home » GR	NUPS » OpenSees Workflows on NeesHub - Pegasus » Wiki » Main Page				
	OpenSees Workflows on NeesHub - Pegasus Wiki New page				
AF					
ATT	Main Page				
	Article Edit Comments History Delete Main Page Index				
pegasus					
	Introduction				
Group Member -	This page documents the effort to run 2 OpenSees workflows through 2 NeesHub/Pegasus on the OSG. The workflow setup is done using				
Overview	Rappture interface on <u>NeesHub</u> , and submitted via Pegasus on the OSG and other resources using the submit command.				
. Members	Rappture Interface				
DD Wiki	The Rappture interface is being developed by Frank 2 McKenna. The purpose is for the user to setup the workflow using the 2 OpenSees				
Resources	executables.				
Discussion	Some screenshots about general properties, record selections, column properties and floor properties are shown below.				
Messages	🗷 Xnest				
Blog	Application:				
¥ Wish List	Opensees ZD Frame Analysis				
Data Sharing					
🖾 Calendar	Graphic + 2 General Properties + G Record Selection + 4 Column Properties + 5 Floor Properties + 6 Simulate				
Discoverability:	Earthquake Records				
Visible					

Credit: Frank McKenna UC Berkeley, NEES, HUBzero



Key Pegasus Concepts

- Workflows are DAGs (or hierarchical DAGs)
 - No loops, no conditional branches
- Pegasus WMS == Pegasus planner (mapper) + DAGMan workflow engine + Condor scheduler
 - The planner does not schedule jobs
- Planning occurs ahead of execution
 - (Except hierarchical workflows)
- Planning converts an abstract workflow into a concrete, executable workflow
 - Planner is like a compiler





Montage Galactic Plane Workflow



John Good (Caltech)

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- Montage Galactic Plane Workflow
 - 18 million input images (~2.5 TB)
 - 900 output images (2.5 GB each, 2.4 TB total)
 - 10.5 million tasks (34,000 CPU hours)
 - Run on Amazon EC2 2013-2014
- Need to support hierarchical workflows and scale





Workflow Application: CyberShake

- What will peak ground motion be over the next 50 years?
 - Used in building codes, insurance, government, planning
 - Answered via Probabilistic Seismic Hazard Analysis (PSHA)
 - Communicated with hazard curves and maps



Hazard curve for downtown LA

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Probability of exceeding 0.1g in 50 yrs



Seismic Hazard Analysis Calculation

- Tensor simulation
 - Create 1.5 billion point mesh with material properties

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- Generate tensors across volume
- Parallel, ~8,000 CPU-hrs



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Post-Processing

- Individual earthquake contributions
 - Get list of earthquakes of interest (~415,000)
 - Simulate seismograms for each earthquake
 - Loosely-coupled, short-running serial jobs
- Combine the levels of shaking with probabilities to produce a hazard curve.





Computational Requirements

	Component	Data	Executions	Cores/exec	Core hours
Tensor Creation	Mesh generation	15 GB	1	320	50
	Tensor simulation	40 GB	2	10,000 CPU 100 GPU	16,000 CPU 1,200 GPU
Post Processing	Tensor extraction	690 GB	6	256	275
	Seismogram synthesis	12 GB	415,000	1	2,300
	Curve generation	1 MB	1	1	< 1
	Total	757 GB	415,000		18,625

This is for **one** location of interest; we wanted to run ~1000





Why Scientific Workflows?

- Large-scale, heterogeneous, high throughput
 - Parallel and many (~415,000) serial tasks
 - Task duration 100 ms 2 hours
- Automation
- Data management
- Error recovery
- Resource provisioning
- Scalable
- System-independent description







CyberShake workflows







Challenge: Resource Provisioning

- In tensor workflow, submit job to remote scheduler
 - GRAM puts jobs in remote queue
 - Runs like a normal batch job
 - Can specify either CPU or GPU nodes
- For post-processing workflow, need high throughput
 - Putting lots of jobs in the batch queue is ill-advised
 - Scheduler isn't designed for heavy job load
 - Scheduler cycle is ~5 minutes
 - Policy limits too
- Solution: Pegasus-mpi-cluster (PMC)



Challenge: Data Management

- Millions of data files
 - Pegasus provides staging
 - symlinks files if possible, transfers files if needed
 - Supports running parts of workflows on separate machines
 - Transfers output back to local archival disk
 - Pegasus registers data products in catalog
 - Cleans up temporary files when no longer needed
- Directory hierarchy to reduce files per directory
- Added automated checks to check integrity
 - Correct number of files, NaN, zero-value checks
 - Included as new jobs in workflow





Challenge: File System Load

- Seismogram tasks cause heavy I/O load
 - Reads an earthquake description
 - Writes a seismogram file
- Reduce reads
 - Generate earthquake description on the fly, from geometry
 - Added memcached to cache rupture geometry
 - Local memory cache on compute node
 - Pegasus-mpi-cluster hook for custom startup script
- Reduce writes
 - Pegasus-mpi-cluster supports "pipe forwarding"
 - Workers write to pipes, master aggregates to fewer files





CyberShake Study 14.2

- Hazard curves for 1144 sites
- 46,720 CPUs + 225 GPUs for 14 days (Blue Waters)
 - Peak of 295,040 CPUs, 1100 GPUs
- 99.8 million tasks executed
 - 81 tasks/sec
 - Only 31,463 jobs in Blue Waters queue
- On average, 26.2 workflows running concurrently
- Managed 830 TB of data
 - 57 TB output files
 - 12.3 TB staged back to local disk (~16M files)
- Workflow tools scale!





