

Scientific Workflows with Pegasus @ CHESS

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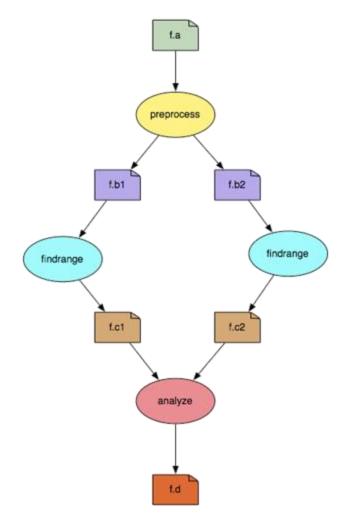
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Scientific Workflows

- An abstraction to express ensemble of complex computational operations
 - Eg: retrieving data from remote storage services, executing applications, and transferring data products to designated storage sites
- A workflow is represented as a directed acyclic graph (DAG)
 - Nodes: tasks or jobs to be executed
 - Edges: depend between the tasks
- Have a monolithic application/experiment?
- The tasks in a scientific workflow can be everything from short serial tasks to very large parallel tasks (MPI for example) surrounded by a large number of small, serial tasks used for pre- and post-processing.
- Find the inherent DAG structure in your application to convert into a workflow





Workflow Challenges Across Domains

- Describe complex workflows in a simple way
- Access distributed, heterogeneous data and resources (heterogeneous interfaces)
- Deal with resources/software that change over time
- Ease of use. Ability to debug and monitor large workflows

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Our Focus

- Separation between workflow description and workflow execution
- Workflow planning and scheduling (scalability, performance)
- Task execution (monitoring, fault tolerance, debugging, web dashboard)
- Provide additional assurances that a scientific workflow is not accidentally or maliciously tampered with during its execution.



Pegasus Workflow Management System

2003

data



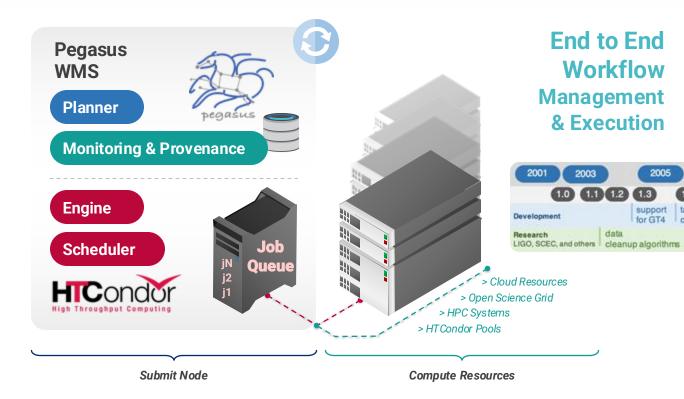
2020

redesign

of APIs

integrity assurance

5.0



- Pegasus in practice
- Laser Interferometer Gravitational Wave Observatory (LIGO) develops large scale analysis pipelines used for gravitational wave detection.
- Southern California Earthquake Center (SCEC) CyberShake project generates hazard maps using hierarchical workflows.
- The XENONnT project uses Pegasus for processing and monte carlo workflows, searching for dark matter
- LIGO observation of colliding black holes

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2005

task

clustering

data

footprint

The XENONnT

detector

support

for GT4

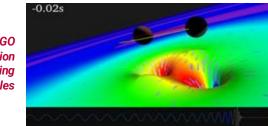
2007

2009

AWS

computing evaluation

cloud



Develop portable scientific workflows in Python, Java, and R

Recover from failures with built-in fault tolerance mechanisms

engine

MPI-based workflow

engine design

Monitor and debug workflow execution via CLI and web-based tools

Regular release schedule incorporating latest research and development

2013

dashboard

data capture

Real time performance

pegasus-lite monitoring

2015

ensemble

manager

Compile workflows to be run on heterogeneous resources

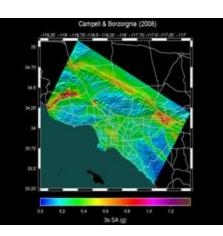
2011

3.0

workflows

2.4

support for hierarchical



2017

metadata

capture

2018

4.8 4.9

support for

containers

data

Hazard map indicating maximum amount of shaking at a particular geographic location generated from SCEC's CyberShake Pegasus workflow









Key Pegasus Concepts

Pegasus WMS == Pegasus planner (mapper) + DAGMan workflow engine + HTCondor scheduler/broker

- Pegasus maps workflows to infrastructure
- DAGMan manages dependencies and reliability
- HTCondor is used as a broker to interface with different schedulers

Workflows are DAGs

- Nodes: jobs, edges: dependencies
- No while loops, no conditional branches
- Jobs are standalone executables

Planning occurs ahead of execution

Planning converts an abstract workflow into a concrete, executable workflow

Planner is like a compiler.





x-pegasus: apiLang: python

createdBy: vahi

pegasus: '5.0'

name: pipeline jobs: - type: job

> name: curl id: ID0000001

arguments:

- o
- pegasus.html

uses:

type: job

name: wc

arguments:

- -1

uses:

id: ID0000002

pegasus.html

- lfn: count.txt

type: output

stageOut: true

registerReplica: true - lfn: pegasus.html type: input jobDependencies: - id: ID0000001 children: - ID0000002

stdout: count.txt

createdOn: 11-19-20T14:57:58Z

http://pegasus.isi.edu

registerReplica: false

- lfn: pegasus.html

type: output
stageOut: false





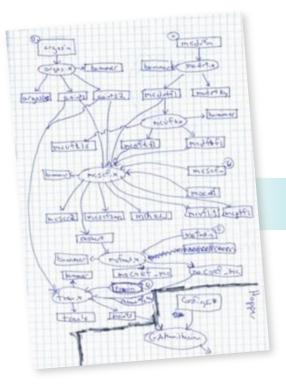
Abstract

Workflow

YAML Formatted

Pegasus provides APIs to generate the Abstract Workflow

#1/usr/bin/env python3



import os	
import logging	
from pathlib import Path	authon"
from argparse import ArgumentParser	ngthon" 🥐
logging.basicConfig(level=logging.DEBUG)	
# Import Pegasus API	
from Pegasus.api import *	
# Create Abstract Workflow	te.
wf = Workflow("pipeline")	🕘 Java
<pre>webpage = File("pegasus.html")</pre>	2
# Create Parent Job	
curl_job = (
Job("curl")	
.add_args("-o", webpage, "http://pegasus.isi.edu")	
.add_outputs(webpage, stage_out=False, register_replica=False)	
)	Jupyter
<pre>count = File("count.txt")</pre>	\sim
	-
# Create Dependent Job	
wc_job = (
Job("wc")	
.add_args(*-1*, webpage)	
.add_inputs(webpage)	
.set_stdout(count, stage_out=True, register_replica=True)	
)	
# Add jobs to the Abstract Workflow	
wf.add_jobs(curl_job, wc_job)	
# Add control flow dependency	
wf.add_dependency(wc_job, parents=[curl_job])	
# Write out the Abstract Workflow	
wf.write()	

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CORNELL HIGH ENERGY

SYNCHROTRON SOURCE

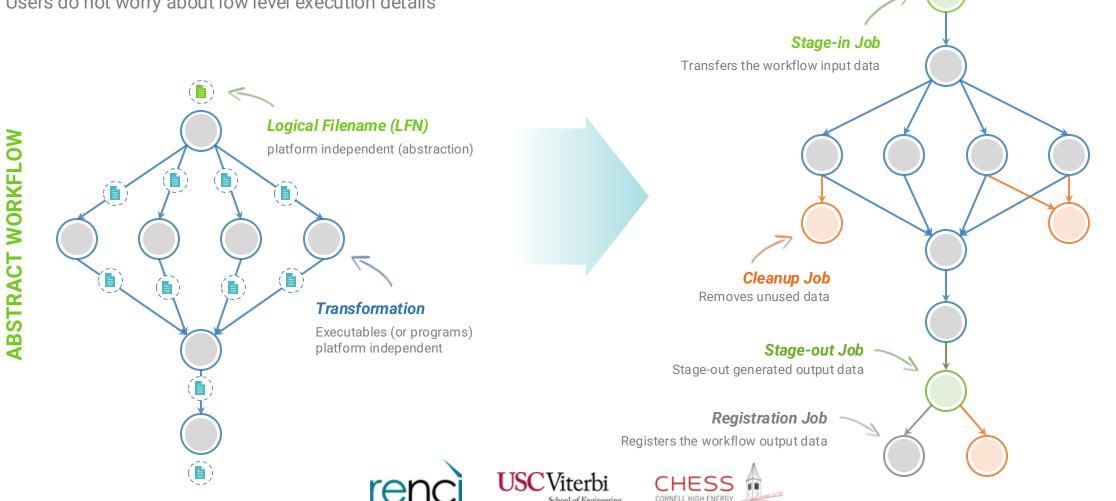


EXECUTABLE WORKFLOW

Input Workflow Specification YAML formatted

Portable Description

Users do not worry about low level execution details



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SYNCHROTRON SOURCE





So, what other information does Pegasus need?



Describes the sites where The workflow jobs are to be executed



Transformation Catalog

Describes all of the executables (called "transformations") used by the workflow



Describes all of the input data stored on external servers





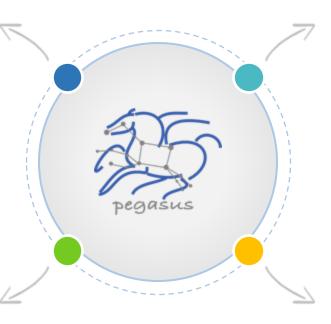
And if a job fails?



detects non-zero exit code output parsing for success or failure message exceeded timeout do not produced expected output files



job generates checkpoint files staging of checkpoint files is automatic on restarts



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helps with transient failures set number of retries per job and run



workflow can be restarted from checkpoint file recover from failures with minimal loss





command-line...

UNRDY READY PRE IN_Q POST DONE FAIL %DONE STATE DAGNAME 14 0 0 1 0 2 0 11.8 Running *split-0.dag \$ pegasus-analyzer pegasus/examples/split/run0001
pegasus-analyzer: initializing...

otal jobs : 7 (100.00%) jobs succeeded : 7 (100.00%) jobs failed : 0 (0.00%) jobs unsubmitted : 0 (0.00%)

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\$ pegasus-statistics -s all pegasus/examples/split/run0001

I	Туре	Succeeded	Failed	Incomplete	Total	Retries	Total+Retries	
I	Tasks	5	0	0	5	0	5	
I	Jobs		0	0		0		
I	Sub-Workflows	0	0	0	0	0	0	

Workflow wall time : 2 mins, 6 secs Workflow cumulative job wall time : 38 secs Cumulative job wall time as seen from submit side : 42 secs Workflow cumulative job badput wall time : Cumulative job badput wall time as seen from submit side : Provenance Data can be Summarized pegasus-statistics Or Used for Debugging pegasus-analyzer





Pegasus Deployment

Workflow Submit Node

- Pegasus WMS
- HTCondor
- Inx201.classe.cornell.edu

One or more Compute Sites

- CHESS Compute Cluster
- Cloud
- OSG

Input Sites

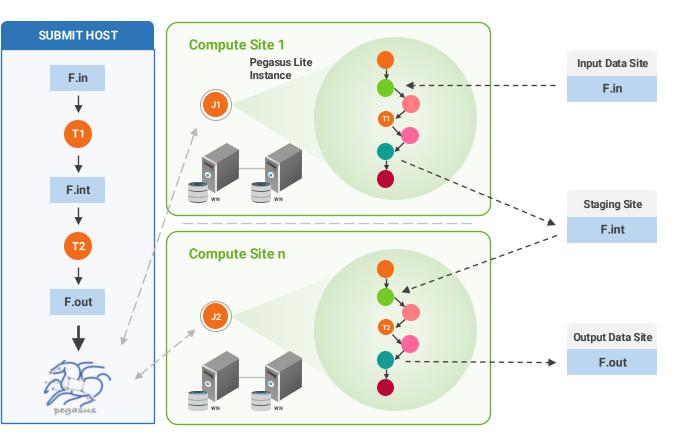
- Host Input Data
- Local Shared Filesystem

Data Staging Site

Coordinate data movement for workflow

Output Site

- Where output data is placed
- Globus Online Endpoint
- Directory on the filesystem









Hands on: Running our first workflow

Submit Host: Logon to *Inx201.classe.cornell.edu*

Reference Materials:

https://xcitecourse.org/theme3/DC101/scientific-workflowmanagement.html#pegasus-workflows

Jump to : <u>Getting Started with Pegasus @ CHESS</u> section.





Data Staging Configurations

HTCondorIO (HTCondor pools, OSG, ...)

- Worker nodes do not share a file system
- Data is pulled / pushed from a staging site, possibly not co-located with the computation
- Staging site is the submit host

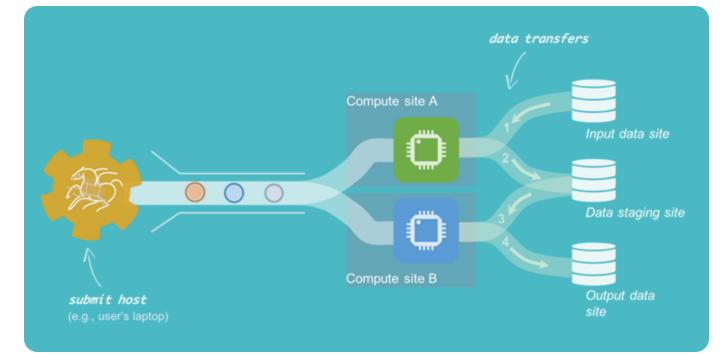
Non-shared File System (clouds, OSG, ...)

- Worker nodes do not share a file system
- Data is pulled / pushed from a staging site, possibly not co-located with the computation
- Staging site is the submit host

Shared File System

(HPC sites, XSEDE, Campus clusters, ...)

I/O is directly against the shared file system







pegasus-transfer

• Directory creation, file removal

If protocol can support it, also used for cleanup

• Two stage transfers

• e.g., GridFTP to S3 = GridFTP to local file, local file to S3

• Parallel transfers

• Automatic retries

Credential management

 Uses the appropriate credential for each site and each protocol (even 3rd party transfers)



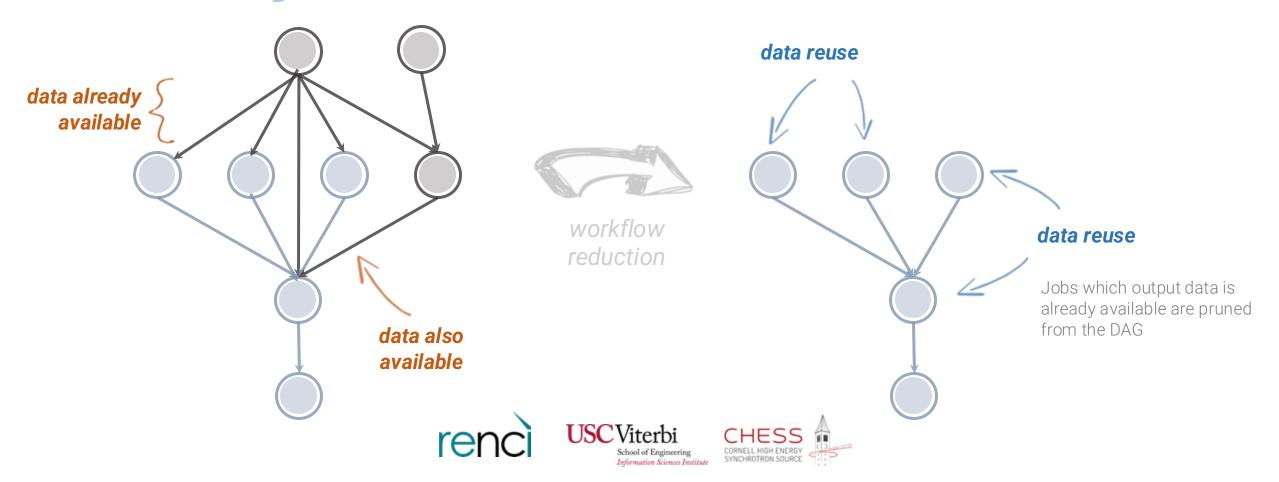
SCP GridFTP Globus iRods Amazon S3 Google Storage SRM FDT Stashcp Rucio Ср ln -s



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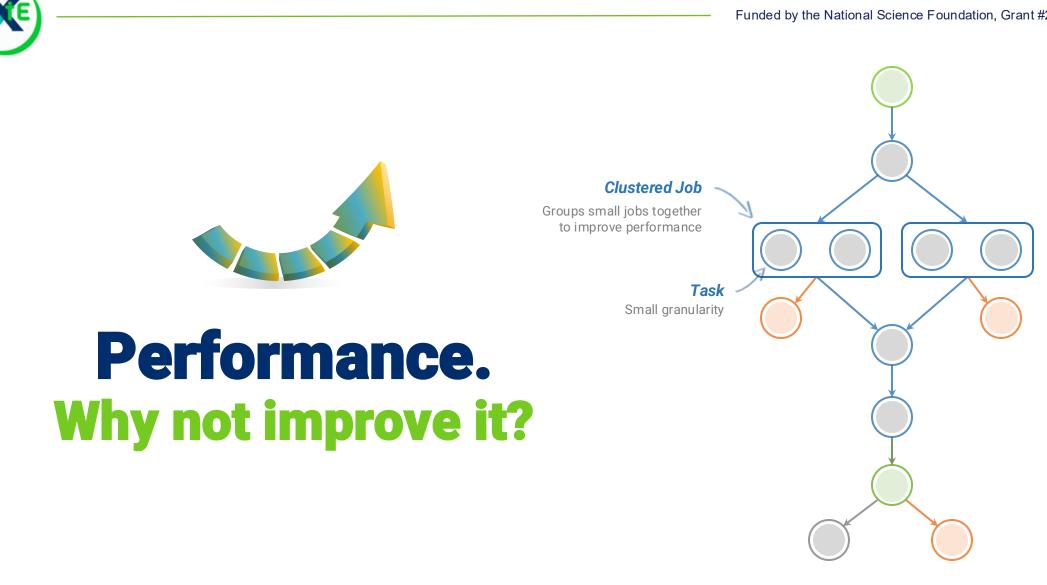


Data Reuse prune jobs if output data already exists













QM2 Beamline

Introductory Video:

https://drive.google.com/file/d/1JCcf66AzHM3XdPb1MLkN hQxcyU85WDtM/view

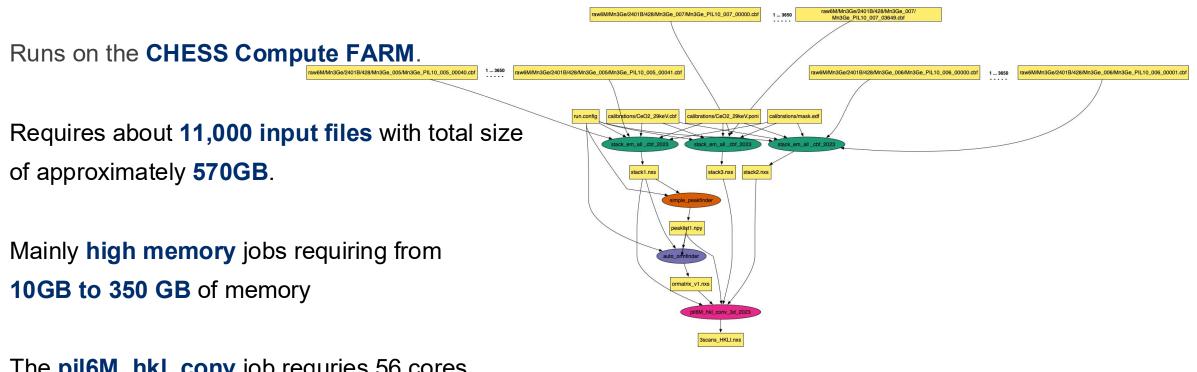






Pegasus CHESS QM Workflow

GitHub Repository: https://github.com/pegasus-isi/chess-qmb-workflow



The **pil6M** hkl conv job requires 56 cores.











XCITE Workflows Module: https://xcitecourse.org/theme3/DC101/scientific-workflowmanagement.html

Website: https://pegasus.isi.edu

Pegasus Users Slack and mailing lists: https://pegasus.isi.edu/contact/

Pegasus Office Hours: <u>https://pegasus.isi.edu/office-hours/</u> <u>https://pegasus.isi.edu</u>



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