



ACCESS Pegasus

A hosted workflow system for ACCESS users, bringing their own capacity

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Pegasus Workflow Management System



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?
 - Find the inherent DAG structure in your application to convert into a workflow







Pegasus Workflow Management





- System End to End Workflow
- Management
- & Execution

- ► Develop portable scientific workflows in Python, Java, and R
- Compile workflows to be run on heterogeneous resources
- Monitor and debug workflow execution via CLI and web-based tools
- Recover from failures with built-in fault tolerance mechanisms
- Regular release schedule incorporating latest research and development

2001 2003		2005		2007		2009		2011		2013	2015	2017	201	8	2020
1.0 1.1	1.2	1.3	1.4	2.0	2.1	2.2 2.3	2.4	3.0 3.1	4.0 4.1	4.2 4.3 4.4	4.5	4.6 4.7	4.8 4	.9	5.0
Development		support for GT4	task clust	ering		support for AWS	hierarc workflo	hical ws	pegasus-lite engine	monitoring dashboard	ensen manag	nble ger	suppor contair	t for ners	redesign of APIs
Research LIGO, SCEC, and others cleanup algorithms		s d	ata otprint	cloud computing evaluation			MPI-based workflow engine design		Real time performance data capture		metadata capture		data ntegri	y assurance	

Visualization of water on nanodiamond spheres from ORNL



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



- Southern California Earthquake Center (SCEC) CyberShake project generates hazard maps using hierarchical workflows.
- Oak Ridge National Lab (ORNL) conducted studies on tRNA and nanodiamonds to improve drug delivery design principles.

LIGO observation of colliding black holes





Pegasus

in practice



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





Input Workflow Specification YAML formatted

Output Workflow

Portable Description

Users do not worry about low level execution details





EXECUTABLE WORKFLOW

Pegasus provides APIs to generate the Abstract Workflow



#!/usr/bin/env python3 import os import logging from pathlib import Path n python from argparse import ArgumentParser logging.basicConfig(level=logging.DEBUG) # --- Import Pegasus API from Pegasus.api import * # --- Create Abstract Workflow ے۔ Java wf = Workflow("pipeline") webpage = File("pegasus.html") # --- 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 count = File("count.txt") # --- 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()





So, what other information does Pegasus need?

Site Catalog

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







Data Reuse prune jobs if output data already exists





Pegasus-transfer

Pegasus' internal data transfer tool with support for a number of different protocols

• Directory creation, file removal

- If protocol can support it, also used for cleanup

• Two stage transfers

• e.g., SCP to S3 = SCP 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)

HTTP webdav SCP GridFTP Globus Online iRods Google Cloud Storage FDT Ср ln -s



Automatic Integrity Checking

Pegasus performs integrity checksums on input files right before a job starts on the remote node.

- For raw inputs, checksums specified in the input replica catalog along with file locations
- All intermediate and output files checksums are generated and tracked within the system.
- Support for sha256 checksums

Support

Job failure is triggered if checksums fail





And if a job fails?



failures with minimal loss



Support

13

Success Stories







421,000 CPU node-hours, 110,000 GPU node-hours

Blue Waters: 673,000 CPU node-hours, 329,000 GPU node-hours





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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)



XENONnT - Dark Matter Search





Two Workflows

Monte Carlo simulations and the main processing pipeline.

- Workflows execute across Open Science Grid (OSG) & European Grid Infrastructure (EGI)
- Rucio for data management
- MongoDB instance to track science runs and data products.



Туре	Succeeded	Failed	Incomplete	Total	Retries	Total+Retries
Tasks	4000	0	0	4000	267	4267
Jobs	4484	0	0	4484	267	4751
Sub-Workflows	0	0	0	0	0	0

Workflow wall time	: 5 hrs, 2 mins
Cumulative job wall time	: 136 days, 9 hrs
Cumulative job wall time as seen from submit side	: 141 days, 16 hrs
Cumulative job badput wall time	: 1 day, 2 hrs
Cumulative job badput wall time as seen from submit side	: 4 days, 20 hrs







8 telescopes: 5 PB of data



50 µas

60 simulations: 35 TB data

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April 11

First images of black hole at the center of the M87 galaxy

Improve constraints on Einstein's theory of general relativity by 500x

480,000 jobs - 2,600,000 core hours

#15 in all OSG projects in last 6 months

#2 in all OSG astronomy projects in the last 6 months

Pegasus-SYMBA Pipeline

Physically accurate synthetic observation data from simulations are keys to develop calibration and imaging algorithms, as well as comparing the observation with theory and interpreting the results.





ACCESS Pegasus



ACCESS Researcher Support Services

- Enable innovative research through equitable and scalable support
- Four tiers of support
- Tools, growing knowledge base
- Match-making with experts
- Student engagement
- **Engagement from community**
- **CSSN** incentives

<u> https://support.access-ci.org</u>





ACCESS ONDEMAND & PEGASUS



ACCESS Support Strategy

Reduce the need for support by simplifying access to resources





& Workflows



INTEGRATED WEB-BASED INTERFACES

Schedule jobs, manage files, create remote visualizations and use a host of other valuable services.



AUTOMATED WORKFLOWS

Simplify complex data workflows on distributed computing resources, such as clusters, grids, and clouds.



ACCESS Pegasus

Leveraging Open OnDemand and Jupyter Notebooks



- 3. Workflows in the tutorials can be run without an allocation.
- 4. For the domain examples, or your own workflows, provision resources under your allocation with HTCondor HPC Annex.

OnDemand version: 3.0.3

OPEN OnDemand

powered by





ACCESS Pegasus

Leveraging Open OnDemand and Jupyter Notebooks



Support



Example Workflows

In addition to tutorial workflows, a set of example workflows are automatically installed into each user account - easy to explore, execute and modify!

- Artificial Intelligence
 - Lung Segmentation
 - Mask Detection
 - Orca Sound
 - LLM + RAG
- Astronomy
 - Montage
- Bioinformatics
 - Alphafold
 - Rosetta
 - VarientCalling





Bring Your Own Capacity

- ACCESS CI consists of allocatable resources
- Motivations
 - Increased / known capacity
 - GPUs
 - Parallel filesystems, I/O
 - Clouds / configurable
- ACCESS Pegasus connects to both shared and BYOC resources

Project Type	Explore	Discover	Accelerate	Maximize
ACCESS Credits*	400,000	1,500,000	3,000,000	Awarded in resource units
Project duration	Supporting grant duration or 12 months	Supporting grant duration or 12 months	Supporting grant duration or 12 months	12 months
Requests accepted	Anytime	Anytime	Anytime	Every 6 months
	Multiple requests allowed	Multiple requests allowed	Multiple requests allowed	1 allowed (some exceptions)
Requirements and review process	Overview	1-page proposal	3-page proposal (max. length)	10-page proposal (max. length)
	Confirmation of eligibility and suitability of requested	Confirmation of eligibility and suitability of requested	Panel merit review	Panel merit review







Shared Capacity - TestPool

Helps users with an ACCESS account but no allocation try the capability

Small amount of compute resources attached to pegasus.access-ci.org

- CPU: 32 cores, 128 GB RAM, 256 GB disk
- GPU: 32 cores, 2 GPUs, 128 GB RAM, 256 GB disk
- hosted on IU Jetstream2, provisioned when needed

Always available, no allocation needed

Can be used for quick turnaround jobs

- workflow development and debugging
- tutorials (not all users might have an allocation at the time of the tutorial)



BYOC - Cloud

- IU JetStream2
- Provided VM image
- Users have to add pegasus.access-ci.org username and token in the cloud-init yaml
- Instances self-terminates when there are no more jobs





Boot Script

This cloud-init ① config describes how to provision the instance. It's provided here to permit specific changes in rare circumstances; please modify it cautiously.

A By editing this it's possible to break various Exosphere features like web desktop, web shell, usage graphs, setup status, etc.

```
#cloud-config
users:
  - default
  - name: exouser
    shell: /bin/bash
    groups: sudo, admin
    sudo: ['ALL=(ALL) NOPASSWD:ALL']{ssh-authorized-
keys}
ssh pwauth: true
package update: true
package_upgrade: {install-os-updates}
packages:
     it(write_files)
bootcmd:

    /opt/ACCESS-Pegasus-Jetstream2/bin/vm-conf alice

aabbcc...
runcmd:
  - echo on > /proc/sys/kernel/printk_devkmsg ||
                                                  true
# Disable console rate limiting for distros that use
kmsq

    sleep 1 # Ensures that console log output from

any previous command completes before the following
command begins
```

BYOC - HTCondor Annex



- Bring your own HPC allocation
- Semi-managed, submits glideins via SSH.
 - A glidein can run multiple user jobs it stays active until no more user jobs are available or until end of life has been reached, whichever comes first.
 - A glidein is partitionable job slots will dynamically be created based on the resource requirements in the user jobs. This means you can fit multiple user jobs on a compute node at the same time.
 - A glidein will only run jobs for the user who started it.
- Documentation: <u>https://htcondor.org/experimental/ospool/byoc/</u>

stampede2

delta cpu

> normal development skx-normal

expanse

compute gpu shared gpu-shared

anvil

wholenode wide shared gpu gpu-debug

bridges2

RM RM-512 RM-shared EM GPU

GPU-shared

path-facility cpu



Shared or BYOC - OSPool



- pegasus.access-ci.org is an OSPool access point, via flocking
- Jobs have to set +ProjectName, which can be an ACCESS allocation (if the user have one), or any registered OSG project name
- No OSDF origin
 - Can use the ACCESS allocated Open Storage Network (OSN), which is a S3 compatible object store



Example AI Workflow - LLM + RAG

LLM RAG (Large Language Model Retrieval-Augmented Generation) is a technique that enhances large language models by incorporating information retrieval mechanisms. It involves retrieving relevant information from a database or document corpus and combining it with the original query to provide additional context. This augmented input is then processed by the large language model to generate more accurate and contextually relevant responses. LLM RAG offers benefits such as improved accuracy, access to up-to-date information, and better contextual understanding, making it useful for applications like question answering, summarization, and conversational AI.

- Container includes LLM
- Inputs: IU Jetstream2 documentation PDF
- Same query twice, once without RAG and one with RAG
- "Please provide a summary of the GPU capabilities of the IU Jetstream2 system. Include the instance flavors, and any details about the GPUs."
- Can execute on GPUS from any of the capacity providers: TestPool, Cloud, HTCondor Annex, OSPool

Summary without RAG:

The IU Jetstream2 system is a high-performance computing resource provided by Indiana University for research purposes. Regarding its GPU capabilities, Jetstream2 offers several instance flavors that include GPUs:

1. `gpu96`: This flavor provides 4 NVIDIA V100 GPUs (with 32GB of memory each) and 128GB of RAM on the host CPU node.

2. `gpu60`: This flavor offers 2 NVIDIA V100 GPUs (each with 32GB of memory) and 64GB of RAM on the host CPU node.

•••

Summary with RAG:

The IU Jetstream2 system has 90 GPU-enabled nodes with four NVIDIA A100 GPUs each. These are divided using NVIDIA virtual GPU (vGPU) to allow allocations to utilize from 1/5th of a GPU to an entire GPU in instances, facilitating use for educational and research workloads requiring varying amounts of GPU processing power.

The Jetstream2 system offers several GPU instance flavors: g3.small, g3.medium, g3.large, and g3.xl. Here's a summary of their specifications:

- g3.small: 4 vCPUs, 15 GB RAM, 60 GB local storage, 20% GPU compute, 20 GB GPU RAM
- g3.medium: 8 vCPUs, 30 GB RAM, 60 GB local storage, 25% GPU compute, 10 GB GPU RAM
- g3.large: 16 vCPUs, 60 GB RAM, 60 GB local storage, 50% GPU compute, 20 GB GPU RAM
- g3.xl: 32 vCPUs, 125 GB RAM, 60 GB local storage, 100% GPU compute, 40 GB GPU RAM



Thank You!

https://support.access-ci.org/tools/pegasus

https://pegasus.isi.edu



Deployment Scenarios

HTCondor Pool

Workflow jobs are submitted to an existing HTCondor pool, and handled directly by HTCondor

Slurm (+HTCondor)

Common setup on campus clusters. Pegasus and HTCondor are installed on a head node. Workflow jobs are submitted to HTCondor, which translates the jobs to Slurm jobs. Good solution for parallel jobs.

Glideins / Pilot Jobs

Variation of HTCondor Pool, but the capacity is dynamically added to the pool by submitting pilot jobs. The pilot jobs will become part of the pool, and will start executing workflow jobs.

Hosted

Larger infrastructure projects might have hosted instances, commonly built on top of either a static pool or glideins. ACCESS Pegasus described later is an example.



AI Workflows - Orcasound

The Ocean Observatories Initiative (OOI), through a network of sensors, supports critical research in ocean science and marine life. Orcasound is a community driven project that leverages hydrophone sensors deployed in three locations in the state of Washington (San Juan Island, Point Bush, and Port Townsend as shown in the figure below) in order to study Orca whales in the Pacific Northwest region.

The workflow processes the hydrophone data of one or more sensors in batches for each timestamp, and converts them to a WAV format. Using the WAV output it creates spectrogram images that are stored in the final output location. Furthermore, using the pretrained Orcasound model, the workflow scans the WAV files to identify potential sounds produced by the orcas.





AI Workflows - Lung Segmentation



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 OPTIMIZATION

MODEL TRAINING



MODEL EVALUATION



Precise detection of the borders of organs and lesions in medical images such as X-rays, CT, or MRI scans is an essential step towards correct diagnosis and treatment planning. We implement a workflow that employs supervised learning techniques to locate lungs on X-ray images. Lung instance segmentation workflow uses Chest X-ray for predicting lung masks from the images using U-Net model.



Automatic Integrity Checking in Pegasus

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Pegasus Container Support



Users can refer to **containers** in the **Transformation Catalog** with their executable preinstalled



Users can **refer** to a **container** they want to **use – Pegasus stages** their executables and containers to the node

- Useful if you want to use a site recommended/standard container image.
- Users are using generic image with executable staging.



Future Plans

- Users can **specify an image buildfile** for their jobs.
- Pegasus will build the Docker image as separate jobs in the executable workflow, export them as a tar file and ship them around







Data Management for Containers





Containers are data too!

Pegasus treats containers as input data dependency

- Staged to compute node if not present
- Docker or Singularity Hub URL's
- Docker Image exported as a TAR file and available at a server, just like any other input dataset

Scaling up for larger workflows

- The image is pulled down as a tar file as part of data stage-in jobs in the workflow
- The exported tar file is then shipped with the workflow and made available to the jobs
- Pricing considerations. You are now charged if you exceed a certain rate of pulls from Hubs

Other Optimizations

- Symlink against existing images on shared file system such as CVMFS
- The exported tar file is then shipped with the workflow and made available to the jobs

