Pegasus WMS : An Introduction and upcoming features

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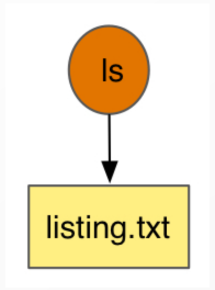
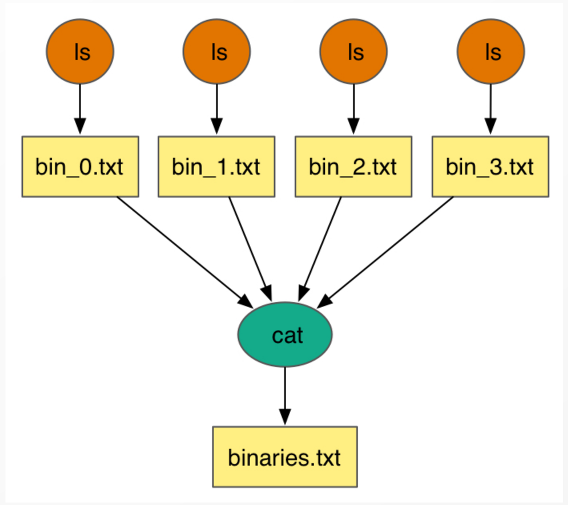
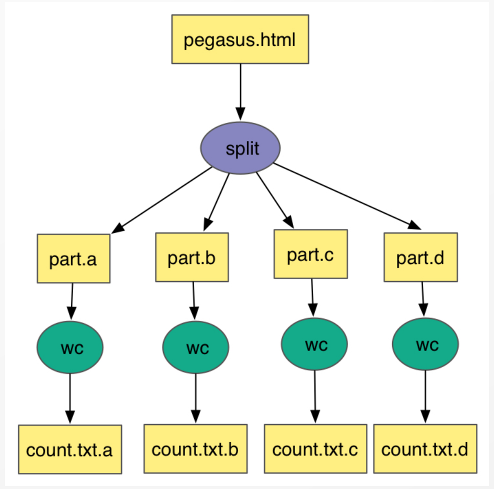
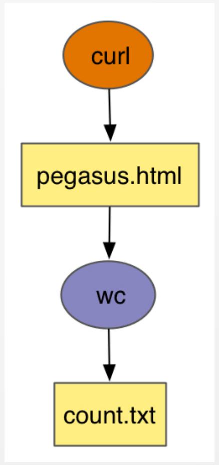
Compute Pipelines – Building Blocks









HTCondor DAGMan

- DAGMan is a **reliable** and a **scalable** workflow executor
  - Sits on top of HTCondor Schedd
  - Can handle very large workflows
- Has useful **reliability** features in-built
  - Automatic **job retries** and **rescue DAG's** (recover from where you left off in case of failures)
- **Throttling** for jobs in a workflow

However, it is still up-to user to figure out

- **Data Management**
  - How do you ship in the small/large amounts data required by your pipeline and protocols to use?
- **How best to leverage different infrastructure setups**
  - OSG has no shared filesystem while XSEDE and your local campus cluster has one!
- **Debug and Monitor Computations.**
  - Correlate data across lots of log files.
  - Need to know what host a job ran on and how it was invoked
- **Restructure Workflows for Improved Performance**
  - Short running tasks?
  - Data placement

Why Pegasus?

Automates complex, multi-stage processing pipelines

Enables parallel, distributed computations

Portable: Describe once; execute multiple times

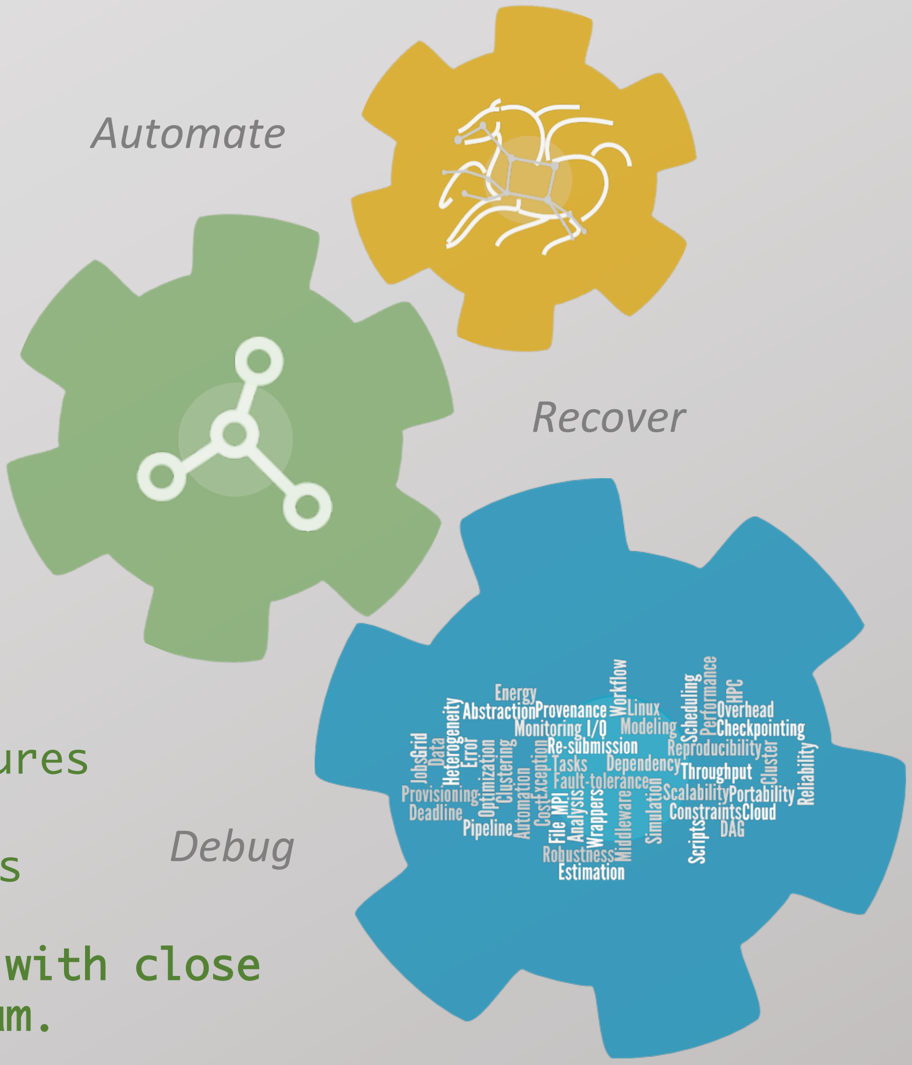
Automatically executes data transfers

Reusable, aids reproducibility

Records how data was produced (provenance)

Provides to tools to handle and debug failures

Keeps track of data and files



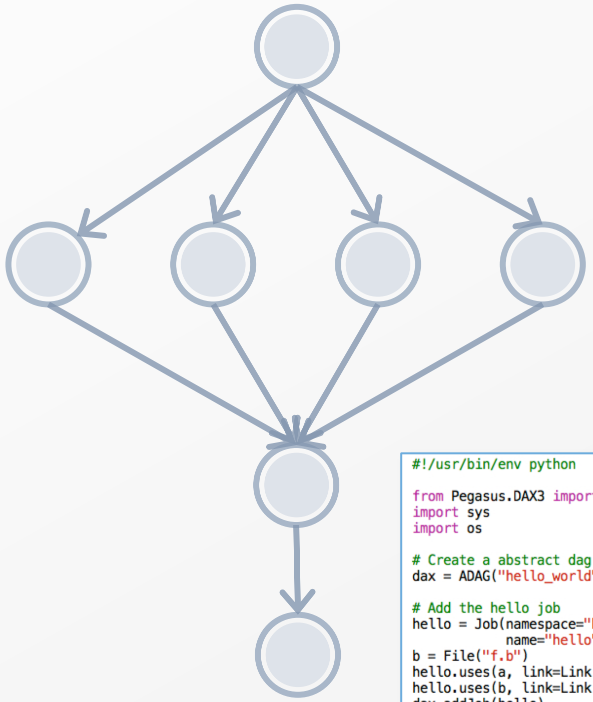
HTCondor
High Throughput Computing

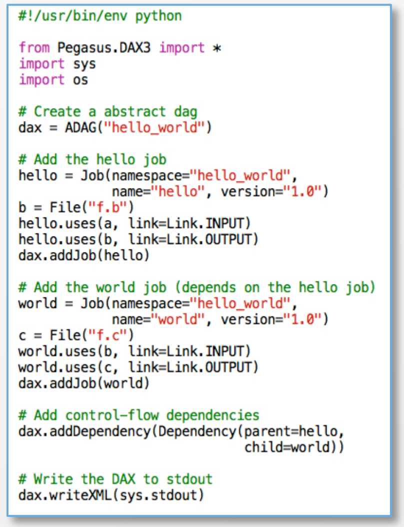
NSF funded project since 2001, with close Collaboration with HTCondor team.

DAX
DAG in XML

Portable Description

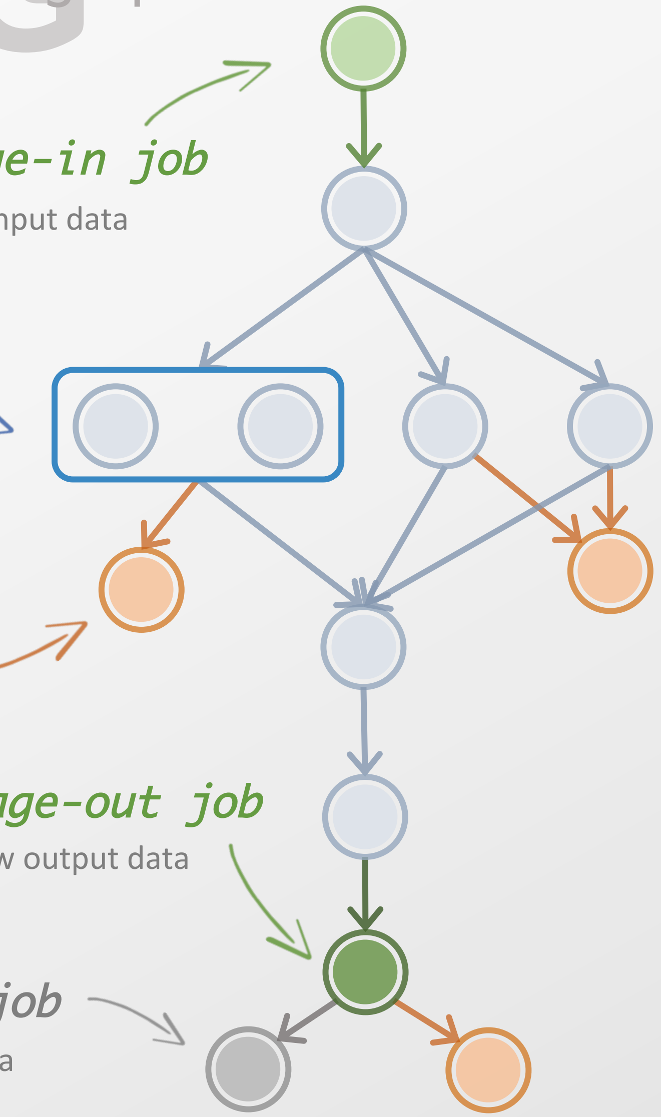
Users don't worry about low level execution details





```
#!/usr/bin/env python

from Pegasus.DAX3 import *
import sys
import os

# Create a abstract dag
dax = ADAG("hello_world")

# Add the hello job
hello = Job(namespace="hello_world",
            name="hello", version="1.0")
b = File("f.b")
hello.uses(a, link=Link.INPUT)
hello.uses(b, link=Link.OUTPUT)
dax.addJob(hello)

# Add the world job (depends on the hello job)
world = Job(namespace="hello_world",
            name="world", version="1.0")
c = File("f.c")
world.uses(b, link=Link.INPUT)
world.uses(c, link=Link.OUTPUT)
dax.addJob(world)

# Add control-flow dependencies
dax.addDependency(Dependency(parent=hello,
                             child=world))

# Write the DAX to stdout
dax.writeXML(sys.stdout)
```

DAG
directed-acyclic graphs



stage-in job

Transfers the workflow input data

clustered job

Groups small jobs together
to improve performance

cleanup job

Removes unused data

stage-out job

Transfers the workflow output data

registration job

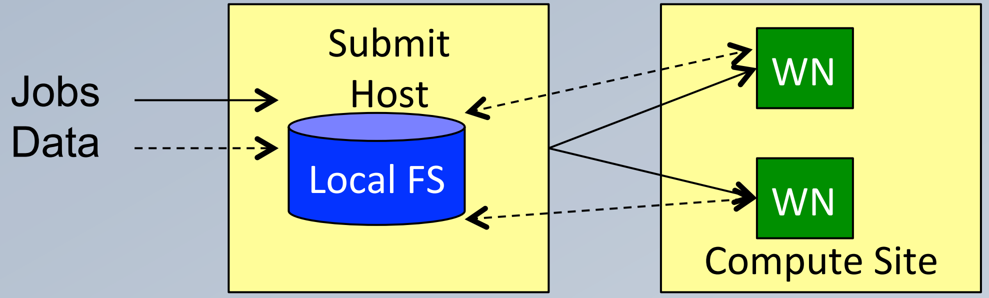
Registers the workflow output data

Pegasus

Data Staging Configurations

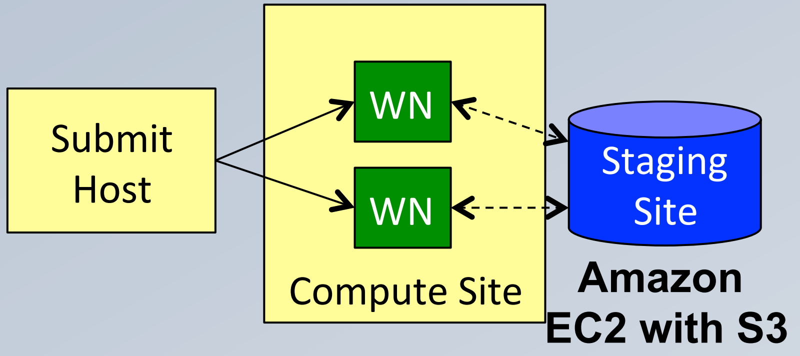
Condor I/O (HTCondor pools, OSG, …)

- Worker nodes do not share a file system
- Data is pulled from / pushed to the submit host via HTCondor file transfers
- Staging site is the submit host



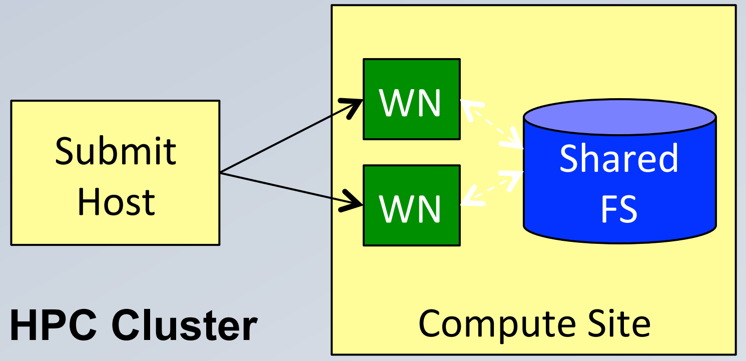
Non-shared File System (clouds, OSG, …)

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



Shared File System (HPC sites, XSEDE, Campus clusters, …)

- I/O is directly against the shared file system

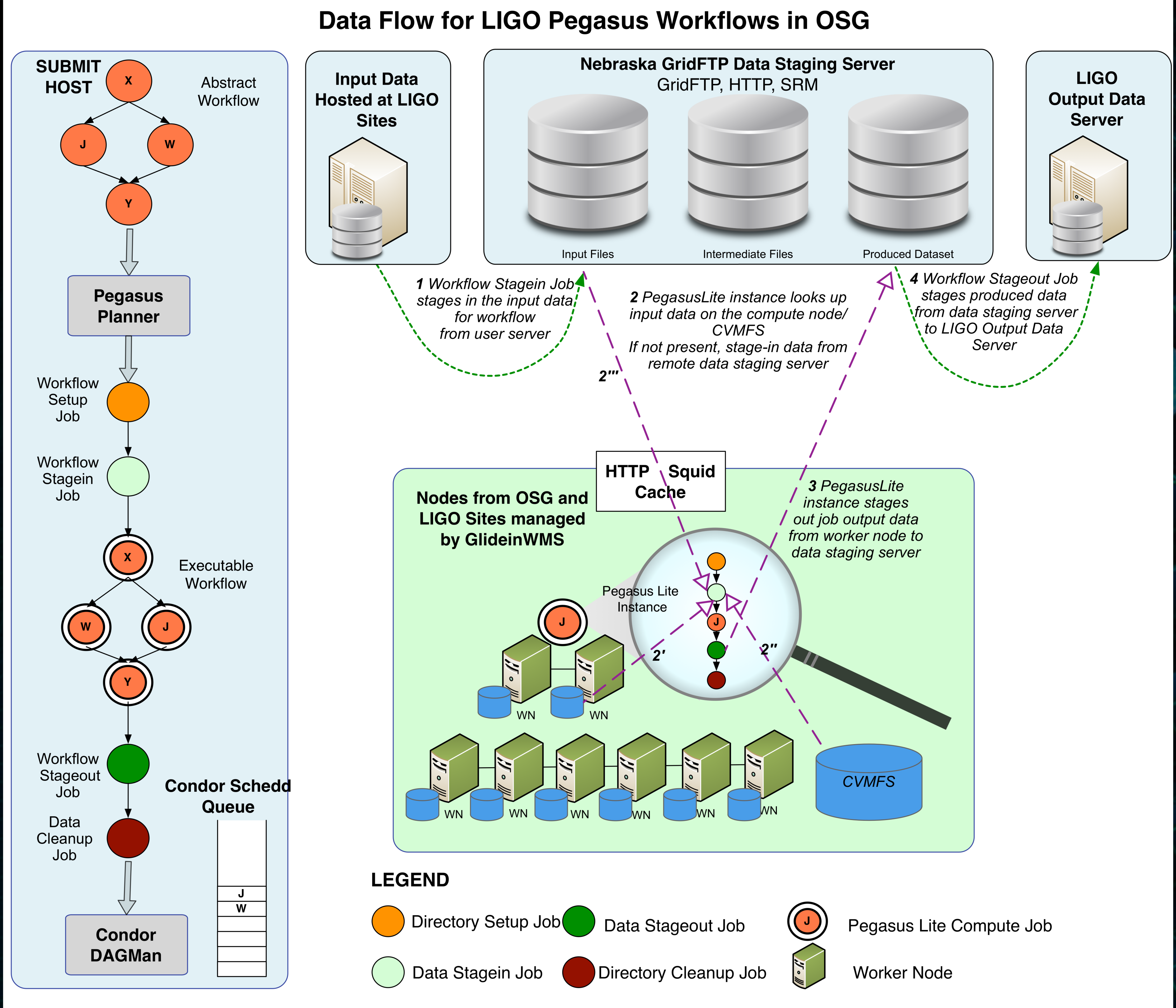


Pegasus Guarantee - Wherever and whenever a job runs it’s inputs will be in the directory where it is launched.

pegasus-transfer

- Pegasus' internal data transfer tool with support for a number of different protocols
- Directory creation, file removal
  - If protocol supports, 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)

HTTP
SCP
GridFTP
Globus Online
iRods
Amazon S3
Google Storage
SRM
FDT
stashcp
cp
ln -s


Data Flow for LIGO Pegasus Workflows in OSG
SUBMIT HOST
Abstract Workflow
X
J
W
Y
Pegasus Planner
Workflow Setup Job
Workflow Stagein Job
Executable Workflow
Workflow Stageout Job
Data Cleanup Job
Condor Schedd Queue
Condor DAGMan
Input Data Hosted at LIGO Sites
Nebraska GridFTP Data Staging Server
GridFTP, HTTP, SRM
Input Files
Intermediate Files
Produced Dataset
LIGO Output Data Server
1 Workflow Stagein Job stages in the input data for workflow from user server
2 PegasusLite instance looks up input data on the compute node/ CVMFS If not present, stage-in data from remote data staging server
4 Workflow Stageout Job stages produced data from data staging server to LIGO Output Data Server
2'''
HTTP Squid Cache
Nodes from OSG and LIGO Sites managed by GlideinWMS
3 PegasusLite instance stages out job output data from worker node to data staging server
Pegasus Lite Instance
2'
2''
WN
CVMFS
LEGEND
Directory Setup Job
Data Stageout Job
Pegasus Lite Compute Job
Data Stagein Job
Directory Cleanup Job
Worker Node

Advanced LIGO – Laser Interferometer Gravitational Wave Observatory

Benefits to LIGO provided by Pegasus- Expanded Computing Horizons

- No longer limited to a single execution resource
  - Non Pegasus LIGO pipelines can often only run on LIGO clusters
  - Input is replicated out of band , in a rigid directory layout.
  - Rely on the shared filesystem to access data.

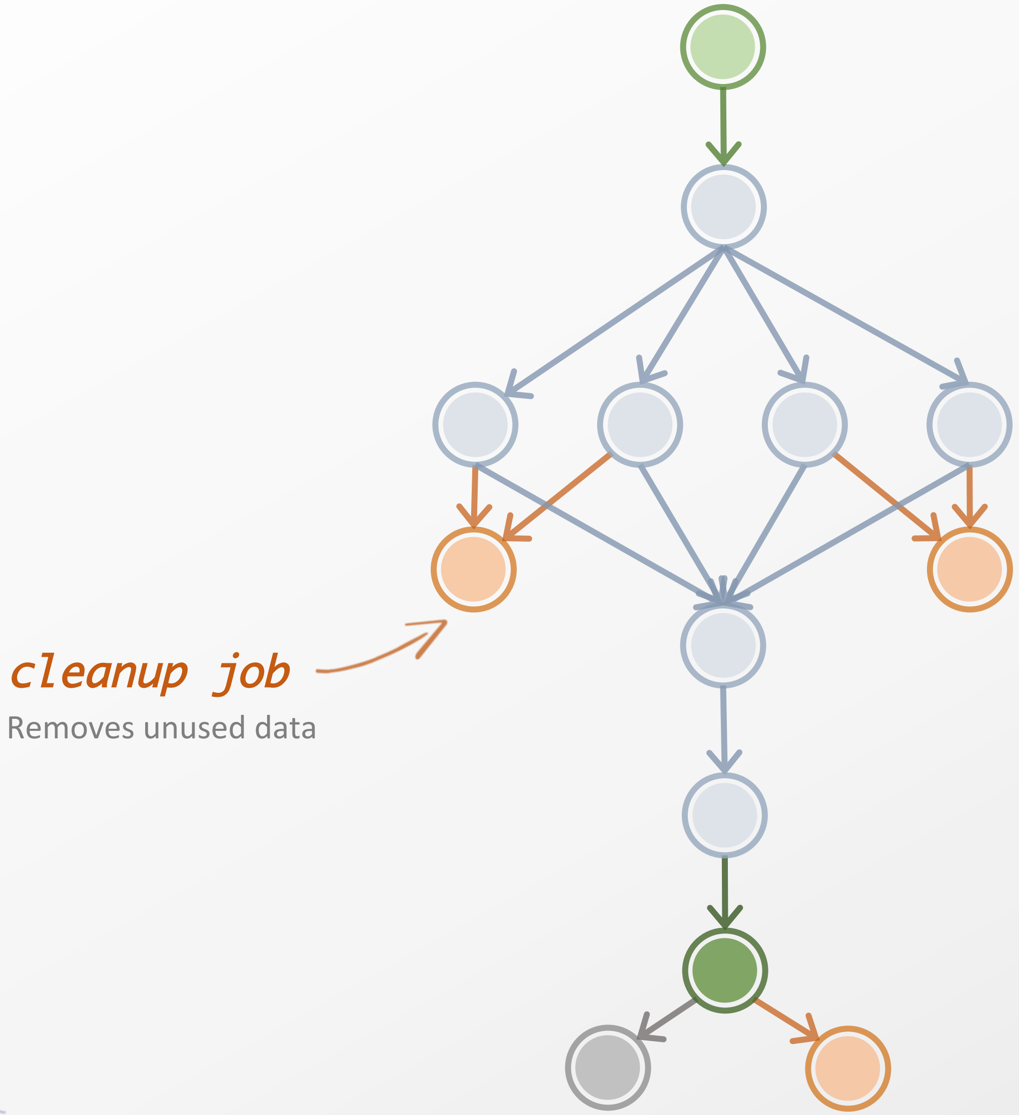
- Pegasus made it possible to leverage Non LDG Computing Resources

  - Open Science Grid
    - Dynamic – Best Effort Resource with no shared filesystem available

  - Large NSF Supercomputing Clusters XSEDE
    - No HTCondor
    - Geared for Large MPI jobs, not thousands of single node jobs
    - LIGO tried to setup XSEDE cluster as a LDG site but mismatch in setup.
    - Pegasus enabled LIGO to use XSEDE without changes at LIGO or at XSEDE

  - VIRGO Resources in Europe
    - Clusters with no shared filesystem and different storage management infrastructure than LDG
    - No HTCondor

Optimizing storage usage…



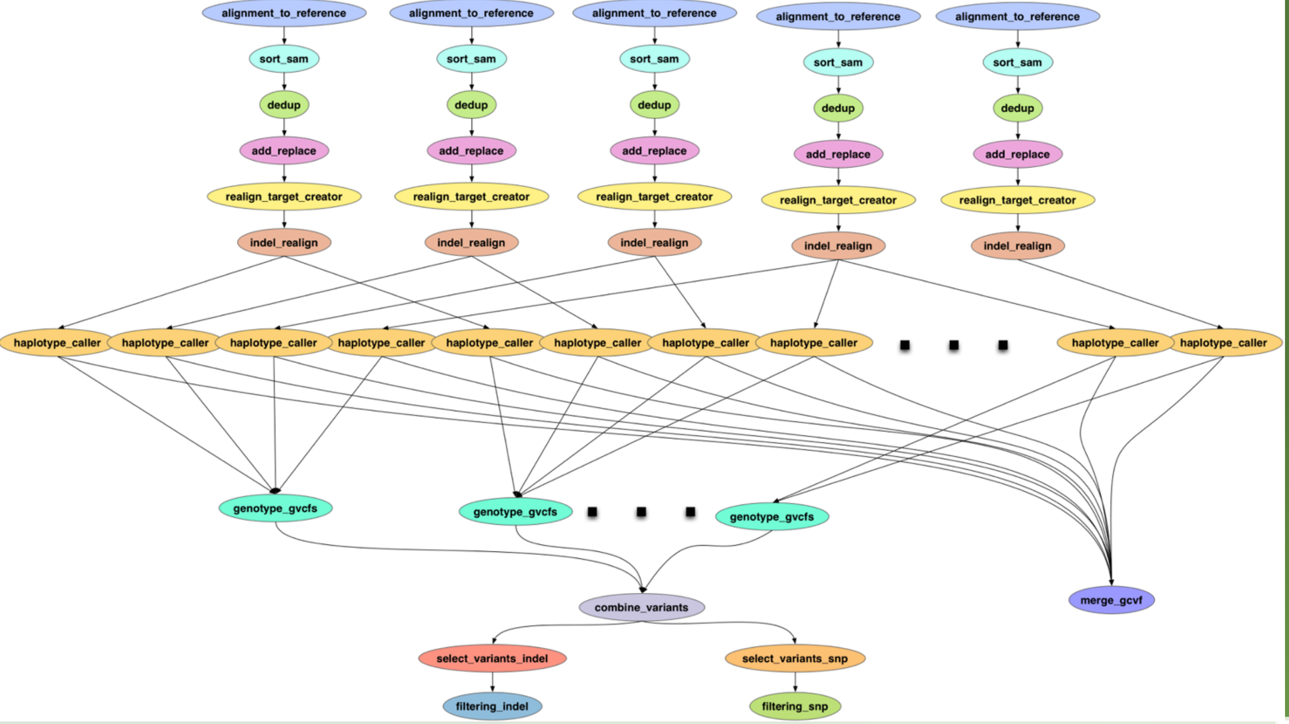
Problem?

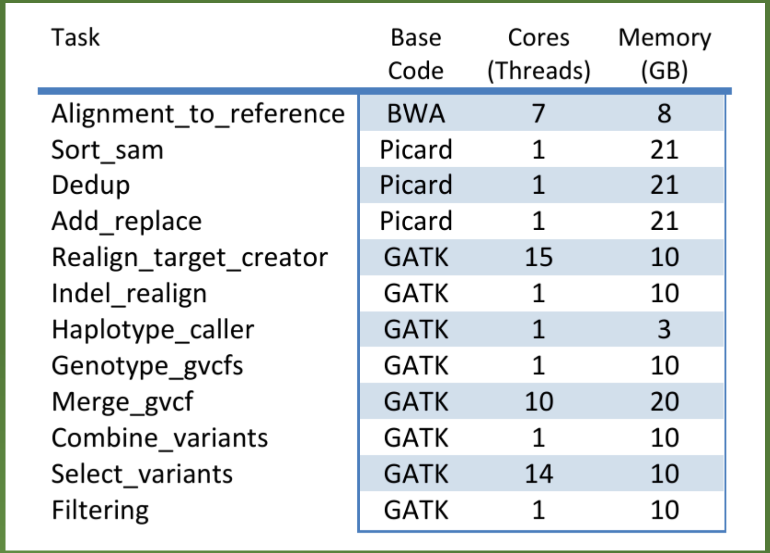
- Users run out of disk space while running workflows

Why does it occur

- Workflows could bring in huge amounts of data
- Data is generated during workflow execution
- Users don't worry about cleaning up after they are done

- Pegasus Solutions
  - Add leaf cleanup nodes to cleanup after workflow finishes.
  - Interleave cleanup nodes
  - Cluster cleanup nodes per level to improve performance





Task	Base Code	Cores (Threads)	Memory (GB)
Alignment_to_reference	BWA	7	8
Sort_sam	Picard	1	21
Dedup	Picard	1	21
Add_replace	Picard	1	21
Realign_target_creator	GATK	15	10
Indel_realign	GATK	1	10
Haplotype_caller	GATK	1	3
Genotype_gvcfs	GATK	1	10
Merge_gvcf	GATK	10	20
Combine_variants	GATK	1	10
Select_variants	GATK	14	10
Filtering	GATK	1	10

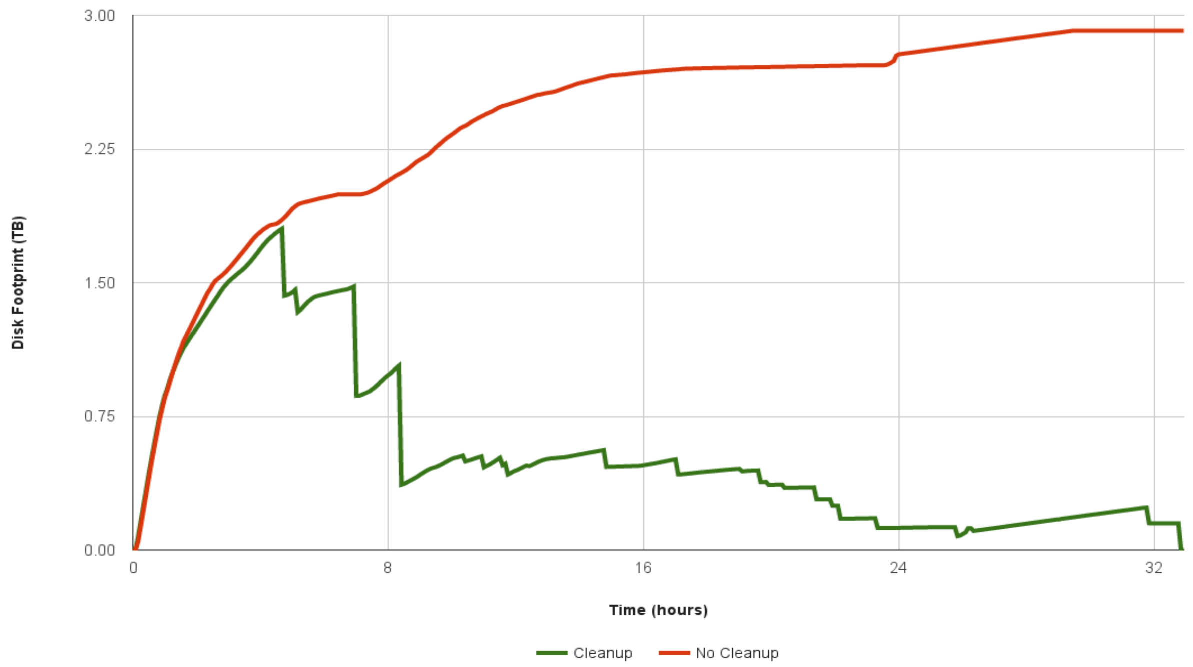
TACC Wrangler as Execution Environment

Flash Based Shared Storage

Switched to glideins (pilot jobs) - Brings in remote compute nodes and joins them to the HTCondor pool on in the submit host - Workflow runs at a finer granularity

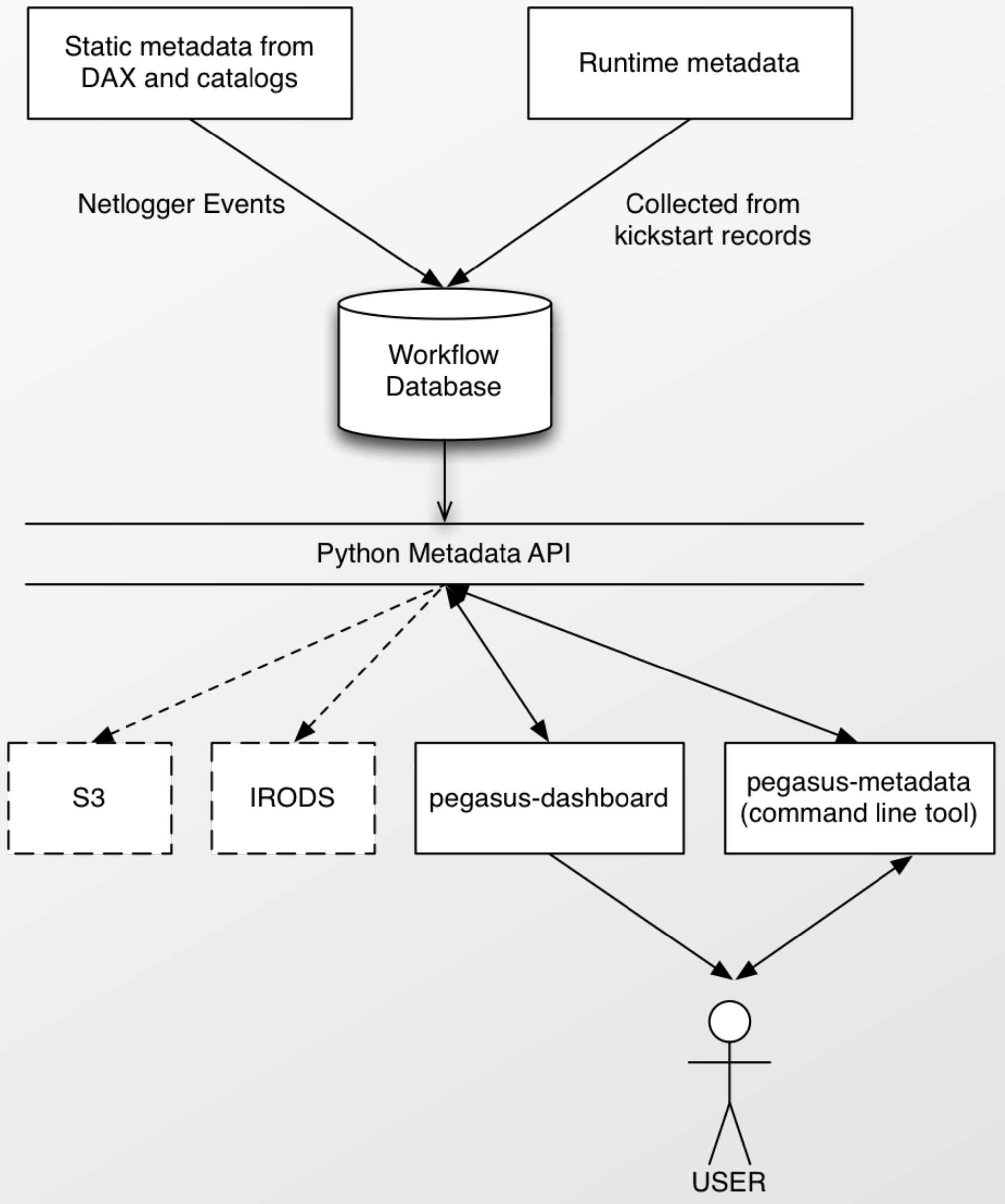
Works well on Wrangler due to more cores and memory per node (48 cores, 128 GB RAM)

Pegasus



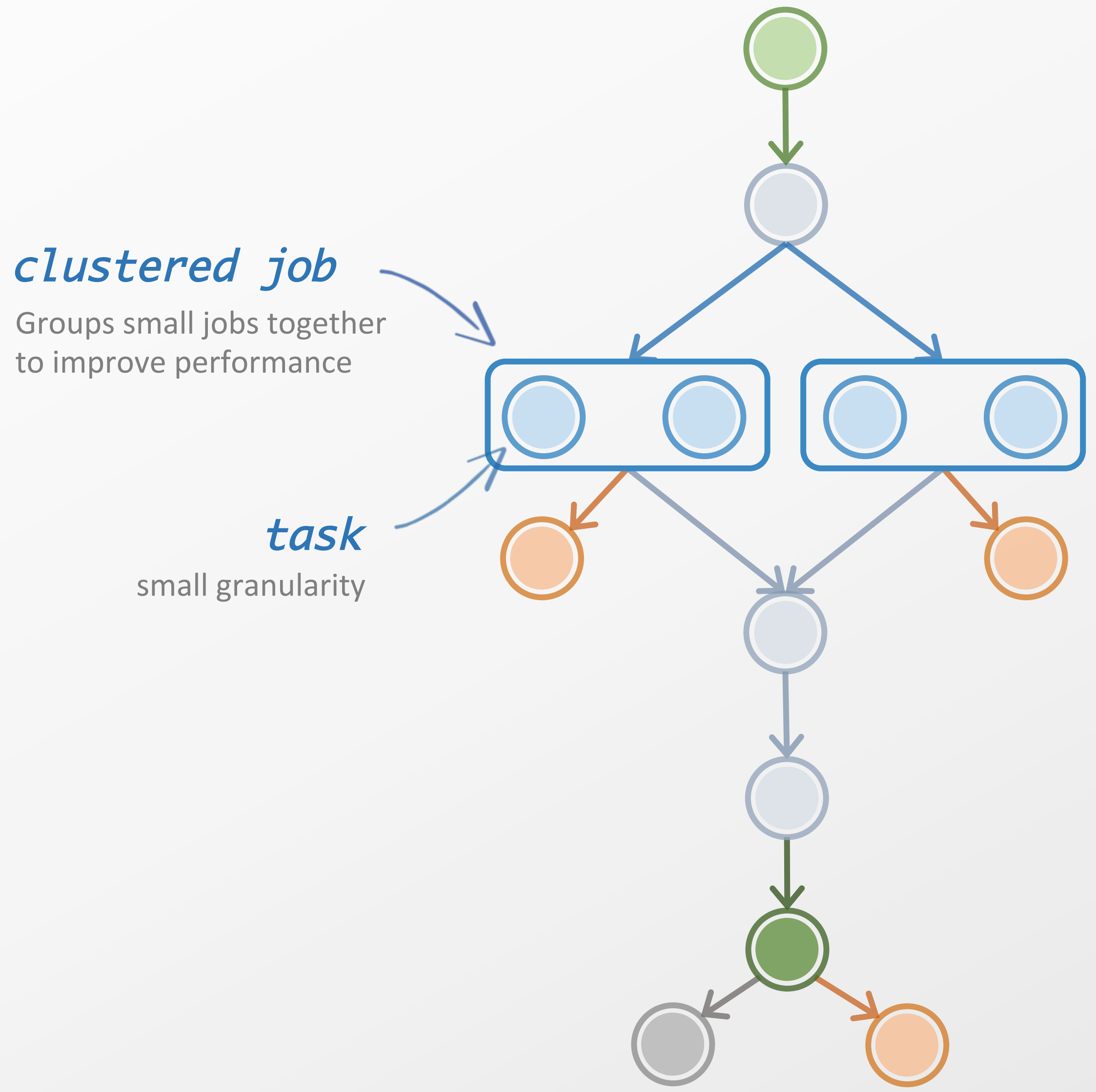
Metadata

- Can associate arbitrary key-value pairs with workflows, jobs, and files
- Data registration
  - Output files get tagged with metadata on registration in the workflow database.
- Static and runtime metadata
  - Static: application parameters
  - Runtime: performance metrics


Static metadata from DAX and catalogs
Runtime metadata
Netlogger Events
Collected from kickstart records
Workflow Database
Python Metadata API
S3
IRODS
pegasus-dashboard
pegasus-metadata (command line tool)
USER

Introduced in Pegasus 4.6

Performance, why not improve it?



Problem?

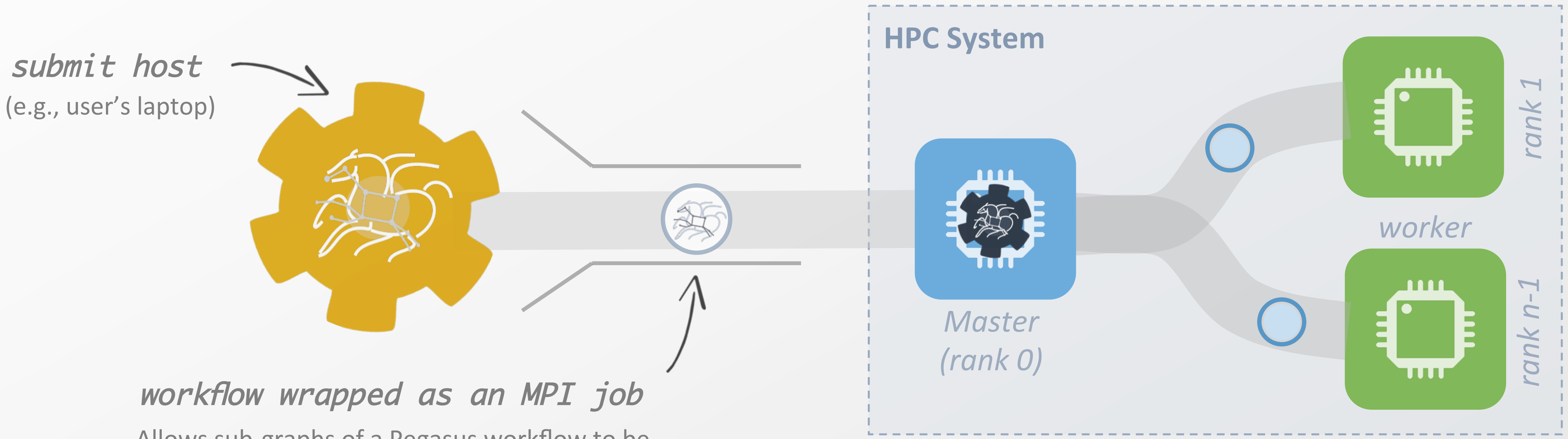
- Users can have short running tasks that increase workflow walltime

Why does it occur

- Each job has a scheduling delay associated with it

- Pegasus Solutions
  - Cluster tasks together resulting in improved performance and better data placement
  - Ability to run clustered tasks as a single MPI job

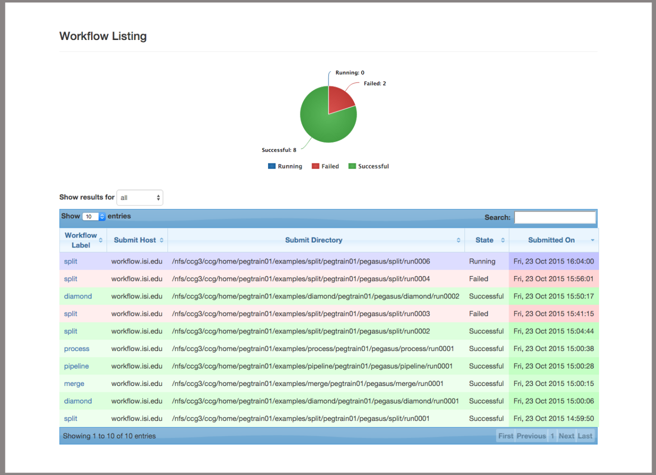
Running fine-grained workflows on HPC systems…



Allows sub-graphs of a Pegasus workflow to be submitted as monolithic jobs to remote resources

Pegasus dashboard

web interface for monitoring and debugging workflows



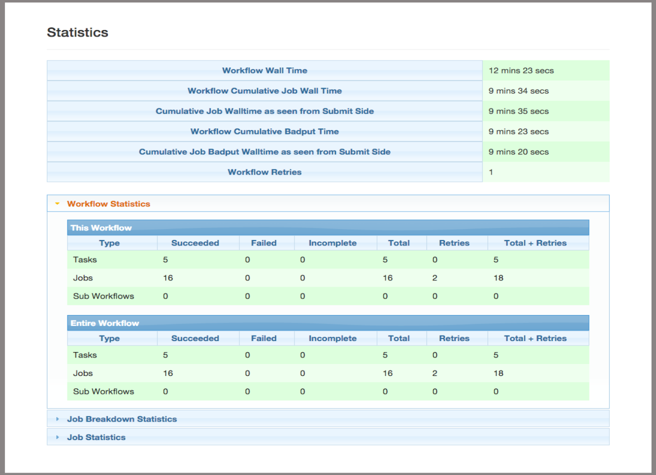
Workflow Listing

Show results for all

Show 10 entries Search:

Workflow Label	Submit Host	Submit Directory	State	Submitted On
split	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/split/pegtrain01/pegasus/split/run0006	Running	Fri, 23 Oct 2015 16:04:00
split	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/split/pegtrain01/pegasus/split/run0004	Failed	Fri, 23 Oct 2015 15:56:01
diamond	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/diamond/pegtrain01/pegasus/diamond/run0002	Successful	Fri, 23 Oct 2015 15:50:17
split	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/split/pegtrain01/pegasus/split/run0003	Failed	Fri, 23 Oct 2015 15:41:15
split	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/split/pegtrain01/pegasus/split/run0002	Successful	Fri, 23 Oct 2015 15:04:44
process	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/process/pegtrain01/pegasus/process/run0001	Successful	Fri, 23 Oct 2015 15:00:38
pipeline	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/pipeline/pegtrain01/pegasus/pipeline/run0001	Successful	Fri, 23 Oct 2015 15:00:28
merge	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/merge/pegtrain01/pegasus/merge/run0001	Successful	Fri, 23 Oct 2015 15:00:15
diamond	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/diamond/pegtrain01/pegasus/diamond/run0001	Successful	Fri, 23 Oct 2015 15:00:06
split	workflow.isi.edu	/nfs/ccg3/ccg/home/pegtrain01/examples/split/pegtrain01/pegasus/split/run0001	Successful	Fri, 23 Oct 2015 14:59:50

Showing 1 to 10 of 10 entries First Previous 1 Next Last



Statistics

Workflow Wall Time	12 mins 23 secs
Workflow Cumulative Job Wall Time	9 mins 34 secs
Cumulative Job Walltime as seen from Submit Side	9 mins 35 secs
Workflow Cumulative Badput Time	9 mins 23 secs
Cumulative Job Badput Walltime as seen from Submit Side	9 mins 20 secs
Workflow Retries	1

Workflow Statistics

This Workflow

Type	Succeeded	Failed	Incomplete	Total	Retries	Total + Retries
Tasks	5	0	0	5	0	5
Jobs	16	0	0	16	2	18
Sub Workflows	0	0	0	0	0	0

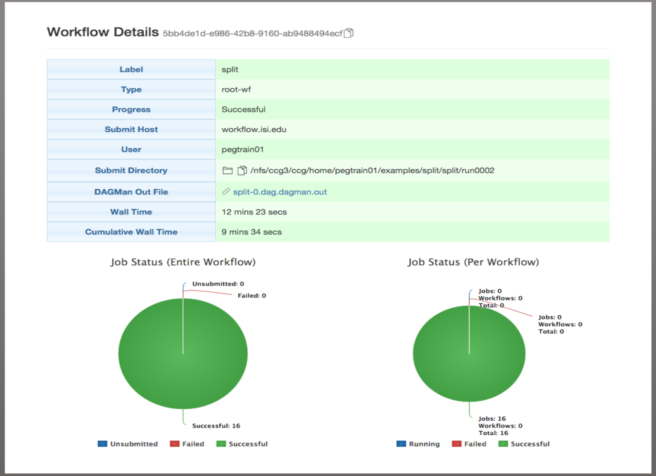
Entire Workflow

Type	Succeeded	Failed	Incomplete	Total	Retries	Total + Retries
Tasks	5	0	0	5	0	5
Jobs	16	0	0	16	2	18
Sub Workflows	0	0	0	0	0	0

Job Breakdown Statistics

Job Statistics

Real-time monitoring of workflow executions. It shows the status of the workflows and jobs, job characteristics, statistics and performance metrics. Provenance data is stored into a relational database.



Workflow Details 5bb4de1d-e986-42b8-9160-ab9488494ecf

Label	split
Type	root-wf
Progress	Successful
Submit Host	workflow.isi.edu
User	pegtrain01
Submit Directory	/nfs/ccg3/ccg/home/pegtrain01/examples/split/split/run0002
DAGMan Out File	split-0.dag.dagman.out
Wall Time	12 mins 23 secs
Cumulative Wall Time	9 mins 34 secs

Real-time Monitoring

Reporting

Debugging

Troubleshooting

RESTful API

But, if you prefer the command-line…

```
$ pegasus-status pegasus/examples/split/run0001
STAT IN_STATE JOB
Run 00:39 split-0 (/home/pegasus/examples/split/run0001)
Idle 00:03  └─split_ID0000001
Summary: 2 Condor jobs total (I:1 R:1)

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...

****************************Summary***************************

Total jobs : 7 (100.00%)
# jobs succeeded : 7 (100.00%)
# jobs failed : 0 (0.00%)
# jobs unsubmitted : 0 (0.00%)
```

```
$ pegasus-statistics -s all pegasus/examples/split/run0001
------------------------------------------------------------------------------
Type           Succeeded Failed Incomplete Total Retries Total+Retries
Tasks               5       0          0       5     0           5
Jobs               17       0          0      17     0          17
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 :
```

…Pegasus provides a set of concise and powerful tools

Upcoming Features

To be released with:

Pegasus 4.8

Automatic Integrity Checking

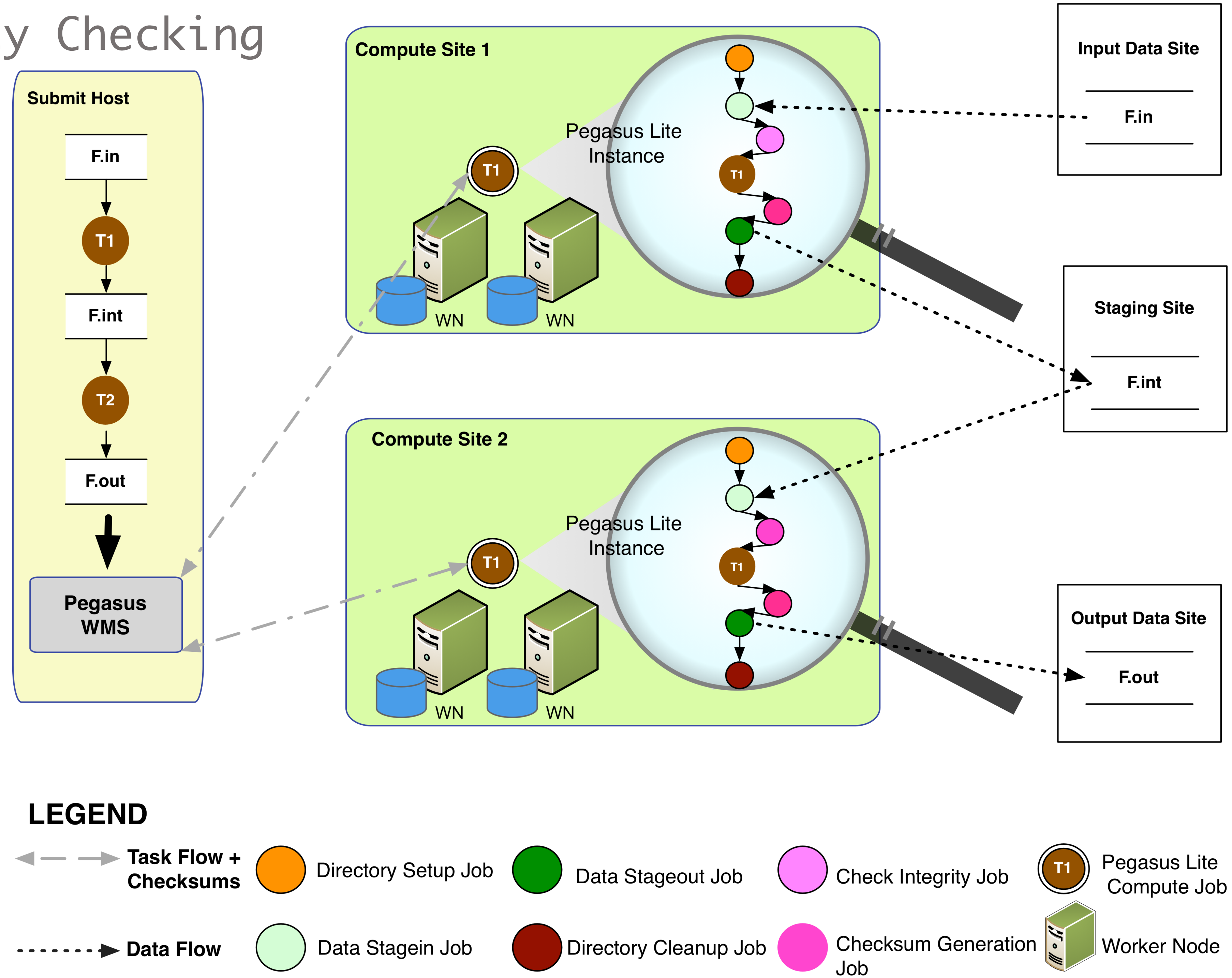
- Pegasus will perform integrity checksums on input files before a job starts on the remote node.

  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

- Failure is triggered if checksums fail



Proposed 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. However, they let Pegasus stage their executable to the node.
  - Useful if you want to use a site recommended/standard container image.
  - Users are using generic image with executable staging.

- Users can specify an image buildfile for their jobs.
  - *Pegasus will build the Docker image as separate jobs in the executable workflow, export them at tar file and ship them around* (planned for 4.8.X)

Plan to support Docker containers initially (due to users preference),
but will be supporting Singularity

docker

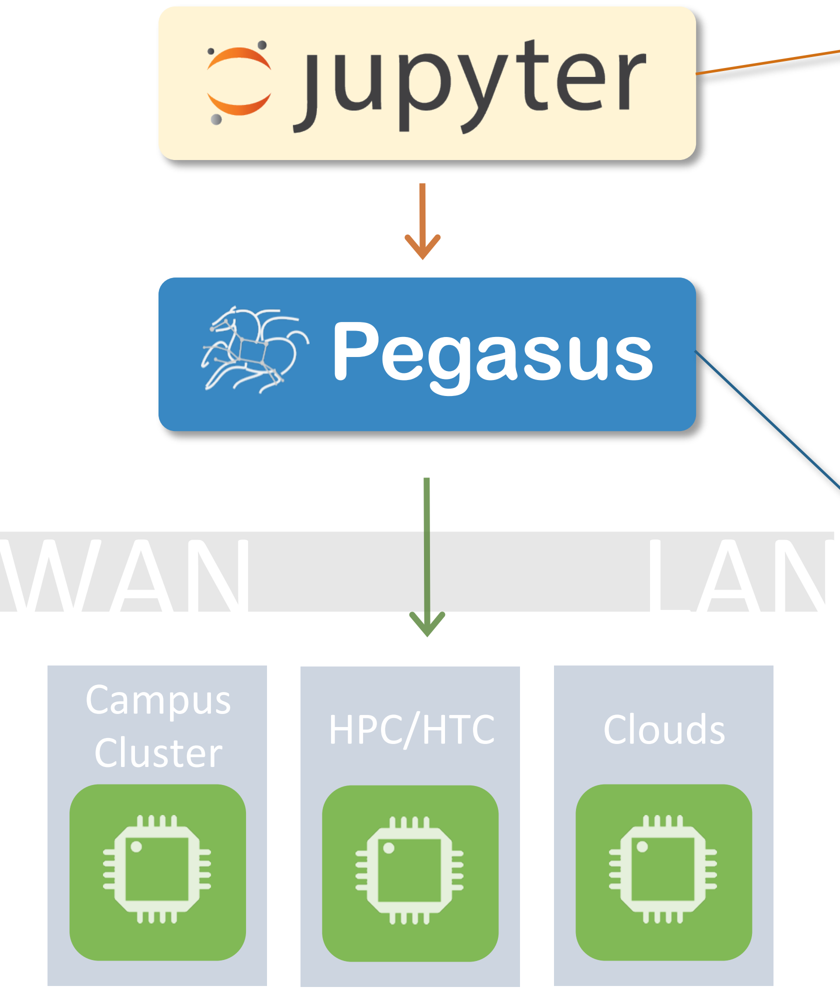
Data Management for Containers

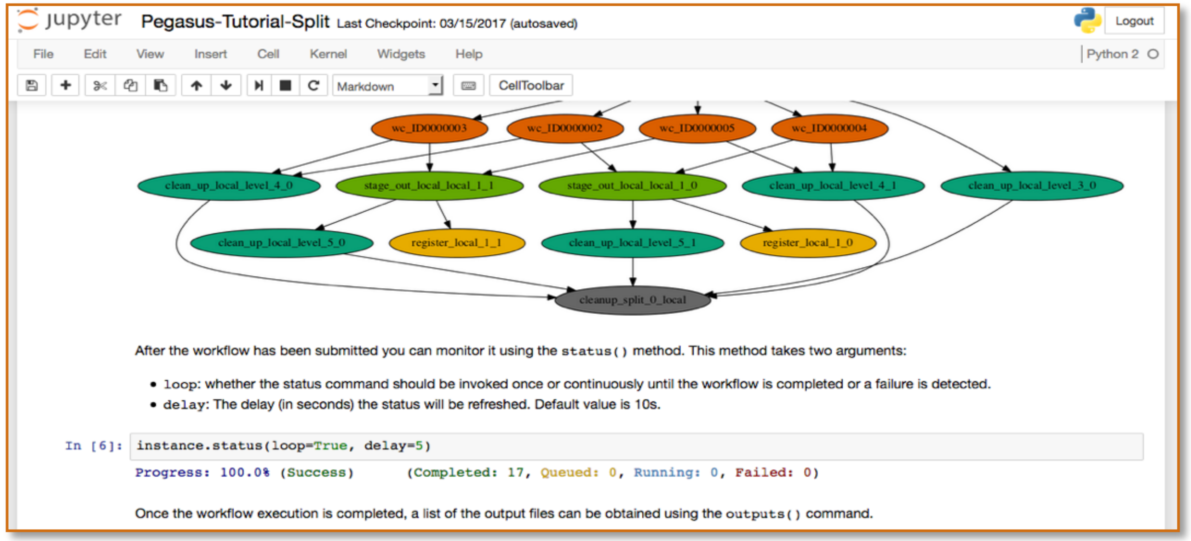
- Users can refer to container images as
  - Docker Hub URL’s
  - Docker Image exported as a TAR file and available at a server , just like any other input dataset.

- We want to avoid hitting Docker Hub repeatedly for large workflows
  - Extend pegasus-transfer to pull image from Docker Hub and then export it as tar file, that can be shipped around in the workflow.

- Ensure pegasus worker package gets installed at runtime inside the user container.

Running Pegasus workflows with Jupyter



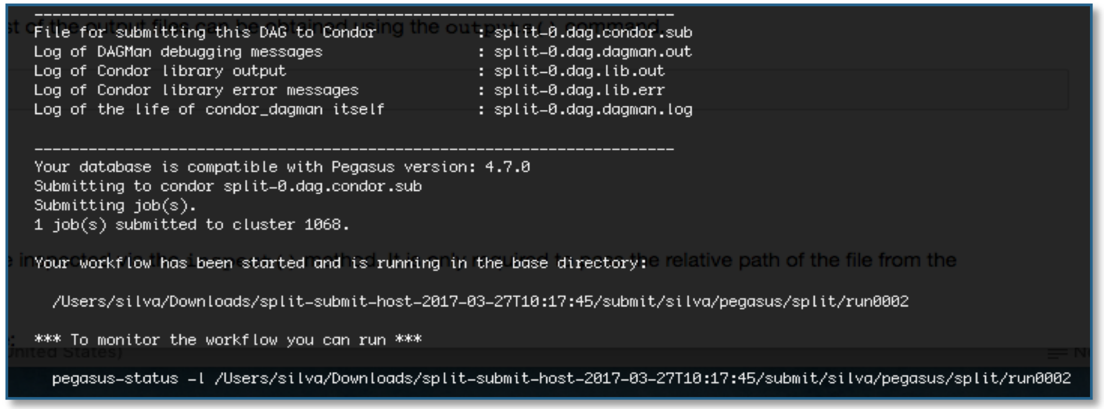


After the workflow has been submitted you can monitor it using the `status()` method. This method takes two arguments:

- `loop`: whether the status command should be invoked once or continuously until the workflow is completed or a failure is detected.
- `delay`: The delay (in seconds) the status will be refreshed. Default value is 10s.

```
In [6]: instance.status(loop=True, delay=5)
        Progress: 100.0% (Success)      (Completed: 17, Queued: 0, Running: 0, Failed: 0)
```

Once the workflow execution is completed, a list of the output files can be obtained using the `outputs()` command.



```
-----------------------------------------------------------------------
File for submitting this DAG to Condor           : split-0.dag.condor.sub
Log of DAGMan debugging messages                 : split-0.dag.dagman.out
Log of Condor library output                     : split-0.dag.lib.out
Log of Condor library error messages             : split-0.dag.lib.err
Log of the life of condor_dagman itself          : split-0.dag.dagman.log

-----------------------------------------------------------------------
Your database is compatible with Pegasus version: 4.7.0
Submitting to condor split-0.dag.condor.sub
Submitting job(s).
1 job(s) submitted to cluster 1068.

Your workflow has been started and is running in the base directory:

  /Users/silva/Downloads/split-submit-host-2017-03-27T10:17:45/submit/silva/pegasus/split/run0002

*** To monitor the workflow you can run ***

  pegasus-status -l /Users/silva/Downloads/split-submit-host-2017-03-27T10:17:45/submit/silva/pegasus/split/run0002
```

Pegasus-Jupyter Python API

```
from Pegasus.jupyter.instance import *
```

importing the API

```
instance = Instance(dax)
```

creating an instance of the DAX

```
# Create an abstract dag
dax = ADAG("split")

# the split job that splits the webpage into smaller chunks
split = Job("split")
split.addArguments("-l","100","-a","1",webpage,"part.")
split.uses(webpage, link=Link.INPUT)
# associate the label with the job. All jobs with same label
# are run with PMC when doing job clustering
split.addProfile( Profile("pegasus","label","p1"))
dax.addJob(split)
```

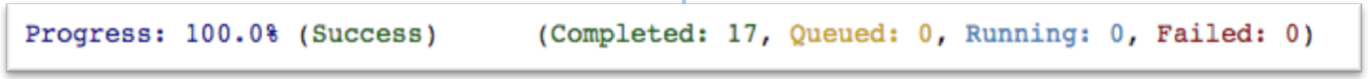
using the Pegasus DAX3 API to write a workflow

```
instance.run(site='condorpool')
```

running a workflow

```
instance.status(loop=True, delay=5)
```

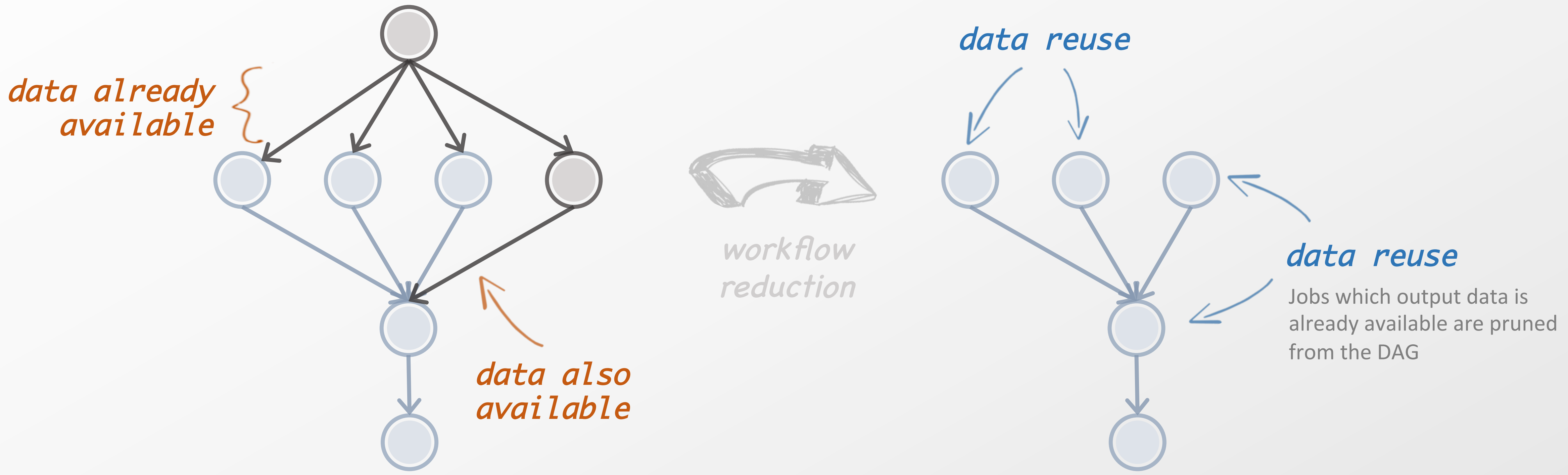
monitoring a workflow execution



```
Progress: 100.0% (Success)        (Completed: 17, Queued: 0, Running: 0, Failed: 0)
```

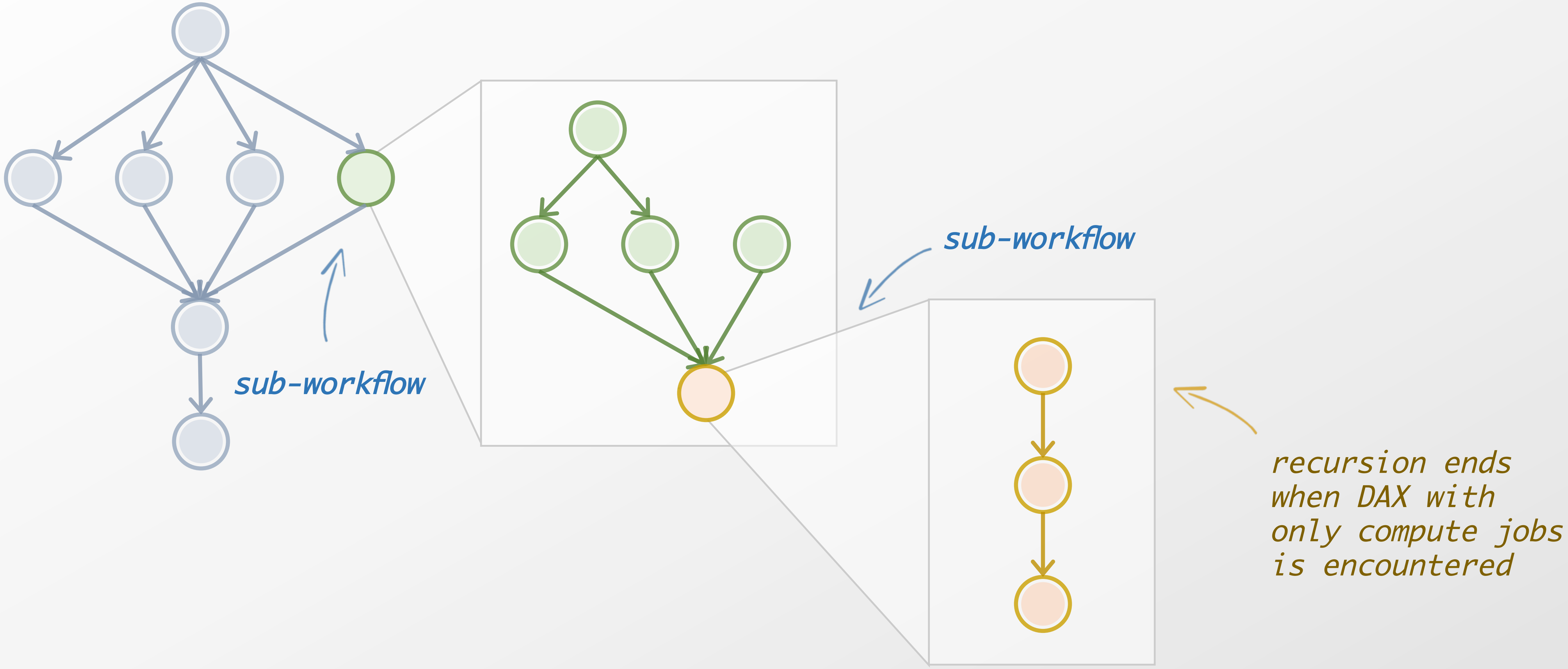
Other Pegasus Capabilities…

What about **data reuse**?



Jobs which output data is already available are pruned from the DAG

Pegasus also handles **large-scale workflows**


sub-workflow
sub-workflow
recursion ends when DAX with only compute jobs is encountered

Pegasus est. 2001

Automate, recover, and debug scientific computations.

Get Started

Pegasus Website

http://pegasus.isi.edu

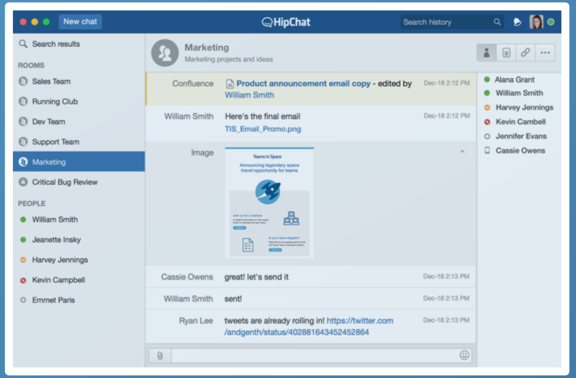
Users Mailing List

pegasus-users@isi.edu

Support

pegasus-support@isi.edu

HipChat



Marketing
Marketing projects and ideas

Confluence	Product announcement email copy - edited by William Smith	Dec-18 2:12 PM
William Smith	Here's the final email TIS_Email_Promo.png	Dec-18 2:12 PM
Image		
Cassie Owens	great! let's send it	Dec-18 2:13 PM
William Smith	sent!	Dec-18 2:13 PM
Ryan Lee	tweets are already rolling in! https://twitter.com/andgenth/status/402881643452452864	Dec-18 2:13 PM

Pegasus est. 2001

Automate, recover, and debug scientific computations.

Thank You

Questions?

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Meet our team

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