



VANDERBILT/JPL COLLABORATION ON A SCALABLE DATA PROCESSING PIPELINE FOR PROTEOMICS

Kirsten Anton, Luca Cinquini, Maureen Colbert, Dan Crichton (PI), Thomas Fuchs, Sean Kelly, Heather Kincaid, Ashish Mahabal, Chris Mattman, Rishi Verma, Paul Zimdars California Institute of Technology & Jet Propulsion Laboratory (NASA)

Matthew Chambers, David Tabb (PI)

Vanderbilt University



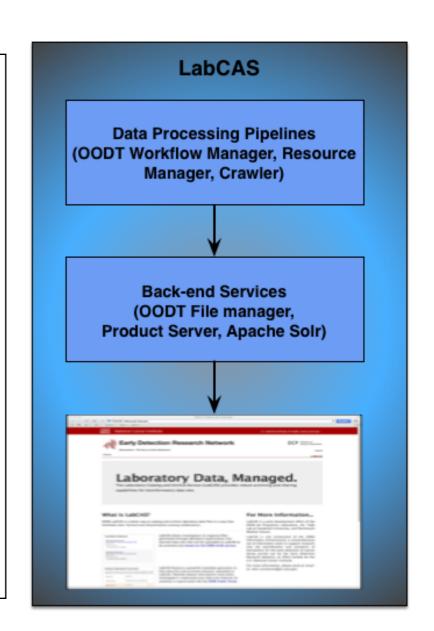
LabCAS: Laboratory Catalog and Archive Services



- <u>LabCAS</u> is an IT integrated environment for managing biomedical laboratory data: generation, publication, documentation, search, discovery and access
- Overall goal: support research for identification and validation of cancer biomarkers for early detection of cancer - focus on early sharing of results before publication
- Developed by JPL in collaboration with EDRN partners (Vanderbilt, Dartmouth, BU...)

Architecture Software Layers

- <u>Data Processing Pipelines</u>:
 - framework for processing laboratory data as integrated workflows, generate data products, and publish them to the archive
- Back-end Services:
 - components for publishing, searching and downloading data products
- Front-end Web Portal:
 - public site for authorized access of data products, as well as UI for execution of data processing pipelines





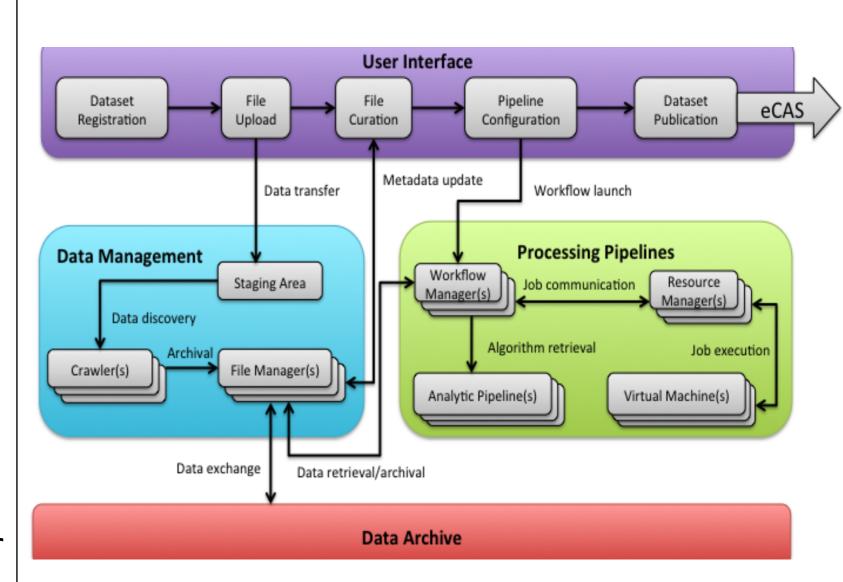
LabCAS: Data Processing Pipelines



Our group is working at establishing a state-of-the-art computing environment at JPL for execution of biology data processing pipelines for the EDRN program.

Goals

- <u>Performance</u>: easy configuration, distributed processing, support multiple pipelines
- Reproducibility: capture
 detailed metadata about
 workflow execution
 ("provenance") so that other
 investigators can reproduce
 and validate the results
- Sharing: publish data and metadata to the web portal for authorized access by other investigators and the general public

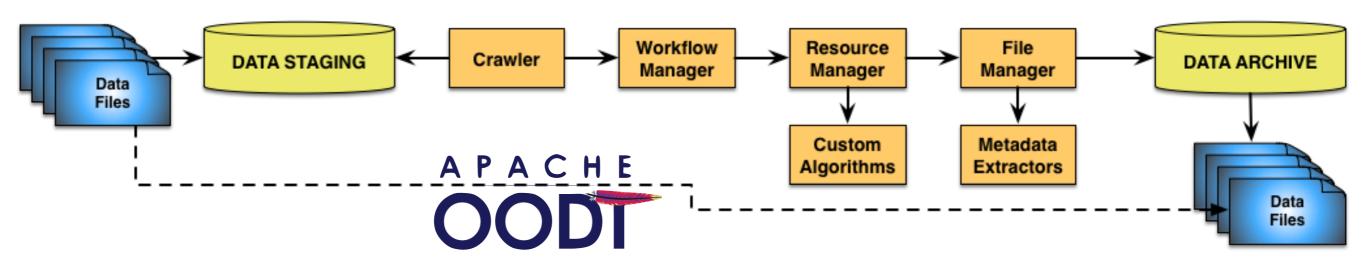




OODT: Object Oriented Data Technology



- Apache OODT is an Open Source framework for management, processing, discovery and access of scientific data
- Modular architecture allows to instantiate and combine different components to realize the most appropriate architecture for a specific data processing environment
 - ▶ <u>File Manager</u>: data access server and metadata catalog. May be backed up by Apache Solr web-enabled, high performance search engine
 - Workflow Manager: general workflow engine for execution of pipelines composed of sequential or parallel tasks
 - ▶ Resource Manager: allocates computing resources for task execution
 - <u>Crawler</u>: service for monitoring data spaces and to trigger ingestion of files into the File Manager or submission of jobs to Workflow Manager

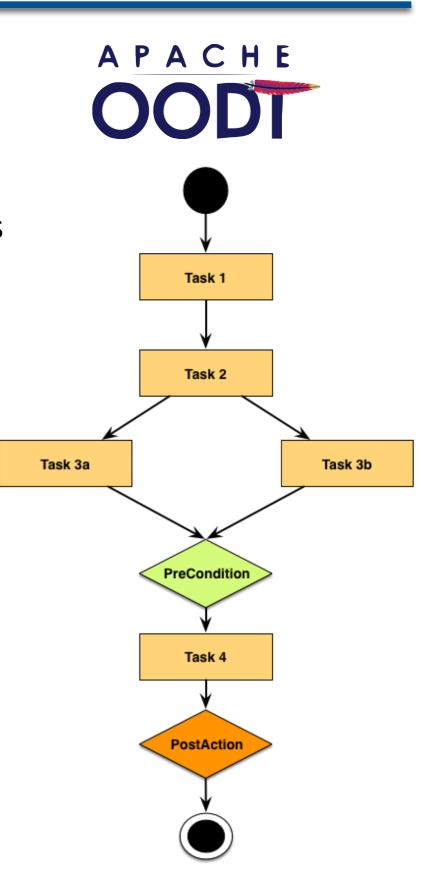




OODT Workflows Features



- Each Workflow Manager server can be configured to execute one or more workflows
- A client starts a workflow by sending an "event" with optional configuration metadata
- Each workflow is composed of an arbitrary number of tasks (sequential or parallel)
- For scalability, tasks can be run on the local Resource
 Manager, or sent to Resource Managers on other servers
- Optional pre-conditions cause task execution to wait until they are satisfied
- Optional post-actions trigger operations when a workflow terminates (on success or failure)
 - ▶ Example: start crawling for products to ingest
- Workflow products are categorized according to custom types:
 - ▶ Specific metadata elements
 - ▶ Specific archive location and versioner
 - ▶ Specific metadata parsers



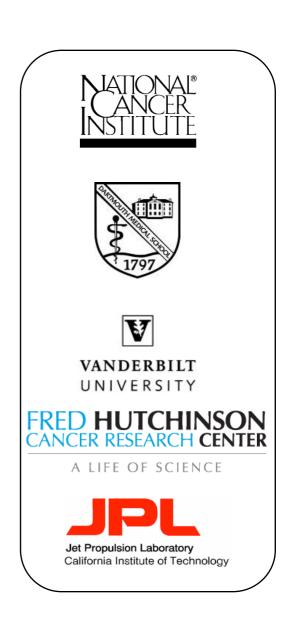


LabCAS Pipelines Collaborations



JPL is collaborating with several EDRN partners to enable their data processing pipelines to be executed within the LabCAS environment:

- First, understand and run the different stages of a pipeline as standalone tasks
- Then, instrument all stages as single runnable OODT workflow
- Vanderbilt University (PI: Dave Tabb)
 - **▶** Proteomics
- Boston University (PI: Marc Lenburg)
 - ▶ Biomarker Discovery
 - ▶ RNA Sequencing
 - ▶ Microarray pre-processing
- University of Washington (PI: Alvin Liu)
 - ▶ Microarray data
- Memorial Sloan Kettering Cancer Center (PI: Gunnar Ratsch)
 - **▶** Genomics
- <u>Cedar-Sinai Medical Center (PI: Beatrice Knudsen)</u>
 - ▶ Pathology
- Cedar-Sinai Medical Center (PI: Michael Freeman)
 - **▶** Proteomics

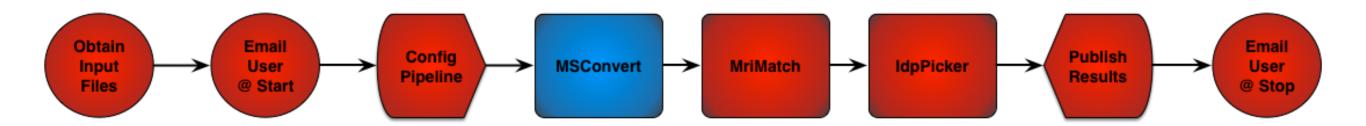




Proteomics Pipeline



- Data processing pipeline for identification and analysis of protein cancer biomarkers in body fluids
- Developed by David Tabb's group at Vanderbilt, part of CPTAC (Clinical Proteomic Tumor Analysis Consortium) activities
- Composed of several programs (part of ProteoWizard suite of open source tools):
 - ▶ MSConvert (pre-processing): conversion of RAW files to mzML format
 - ▶ MyriMatch (database search): searches sample data for peptides (.pepXML)
 - ▶ IdpPicker (filtering: IdPQonvert+IdPAssemble+IdPQuery): generates protein identification reports
- Instrumented as single OODT workflow executing a sequence of 13 tasks
- First test case at JPL consisted in running the pipeline on a medium-size pancreatic dataset composed of 675 input files of approx. 200MB each (total: 1.35TB)

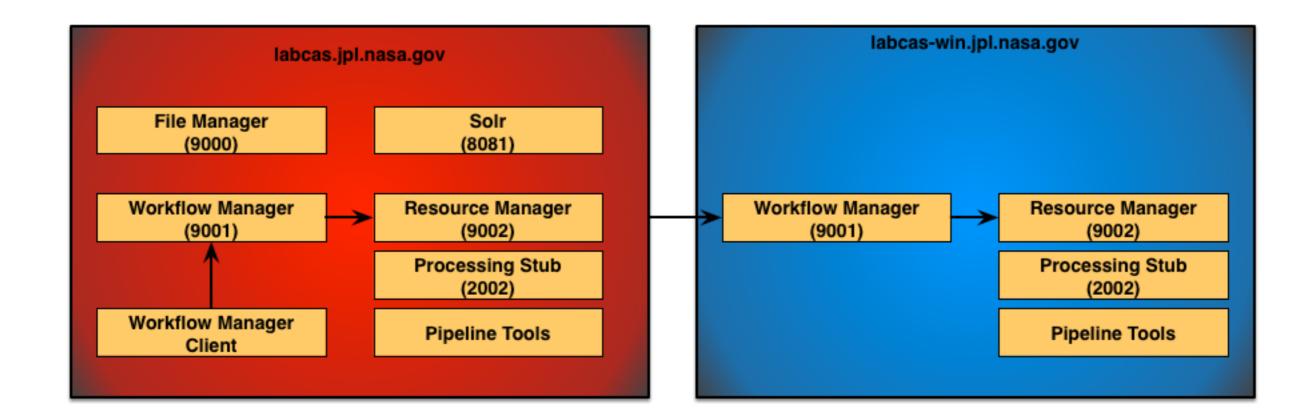




Proteomics Pipeline @ JPL: 1st attempt



- Use system of 1 Linux + 1 Windows servers (2 cores each)
- Each task setup to run over all files sequentially
- Tasks 1-7 (up to MSConvert) completed in about 1 week on Windows server
- Task 8 (MyriMatch) started on Linux server, stopped after estimating 40+ days before completion (approximately 1.5 hours for each file)

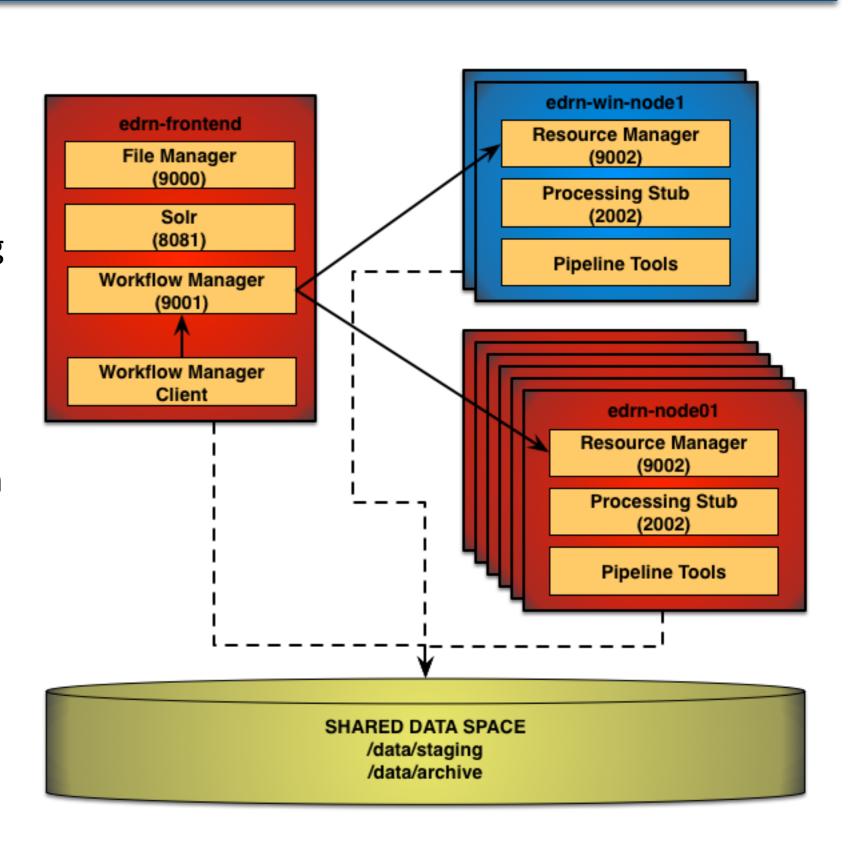




New EDRN Cluster Environment @ JPL



- System of 21 Virtual Machines (6 cores each)
 - ▶ 1 front-end server hosting common services
 - ▶ 18 Linux back-end processing nodes
 - 2 Windows back-end processing nodes
 - ▶ 10TB shared storage
- Automatic software replication from 1 processing node to all the others
- Execution of commands from front-end node to all the backend nodes
 - Start/stop OODT services on all node simultaneously





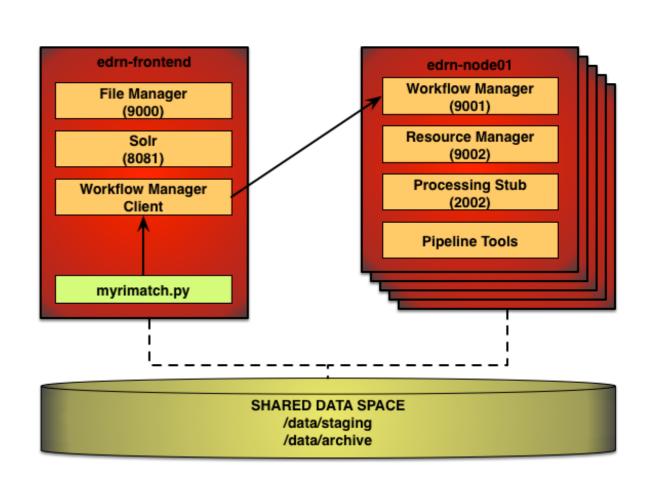
Proteomics Pipeline @ JPL : 2nd attempt



- Use new JPL/EDRN cluster
- MyriMatch step was identified as bottle-neck in Proteomics pipeline
- Because MyriMatch only processes 1 mzML file at a time, this step can be fully parallelized!
- •Using mzML files already produced by the old pipeline, sub-workflow was setup to run MyriMatch in parallel over all available Linux nodes

Results

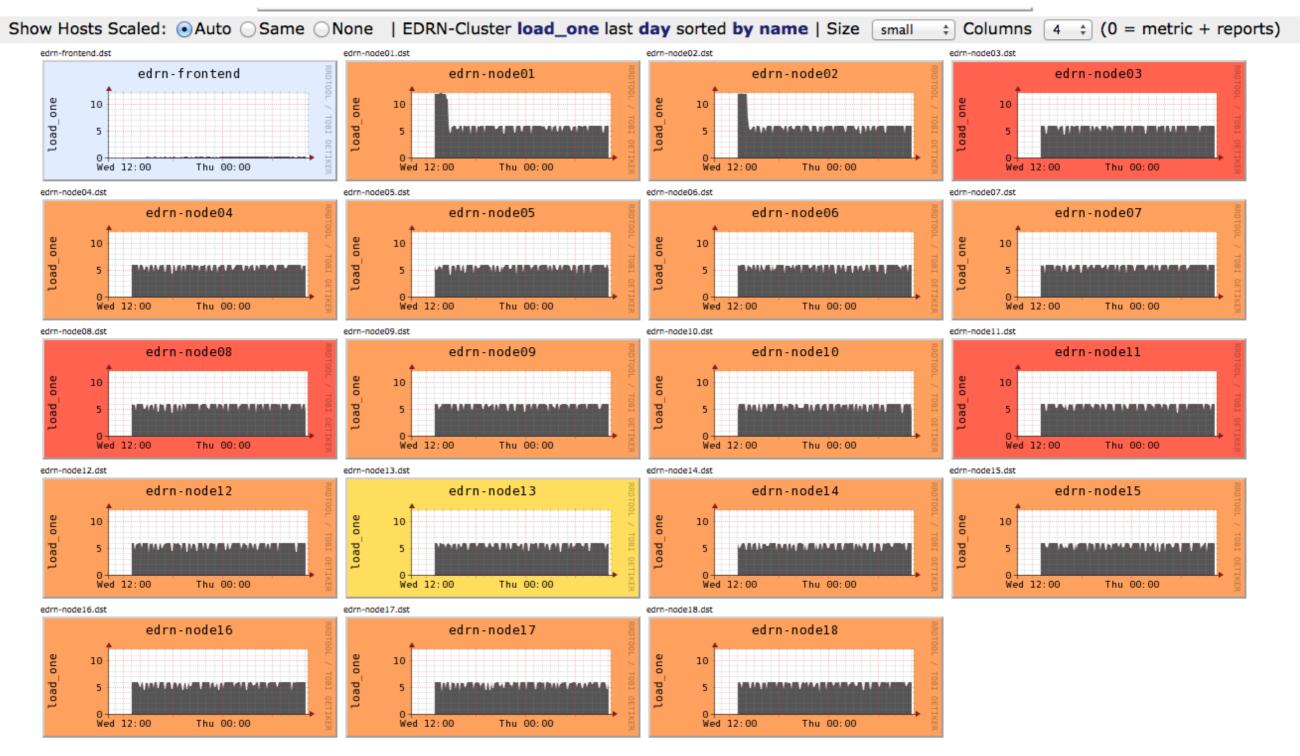
- Single MyriMatch job completes in approx. 30min (instead of 90+ min)
- Full MyriMatch sub-workflow completed in approx. 1 day (for all 675 files)!
- Performance improvements:
 - ▶ Factor of 3 from using more powerful hardware
 - ▶ Factor of 18 from parallelizing a single step onto multiple processing nodes





JPL/EDRN Cluster Usage (1 day)

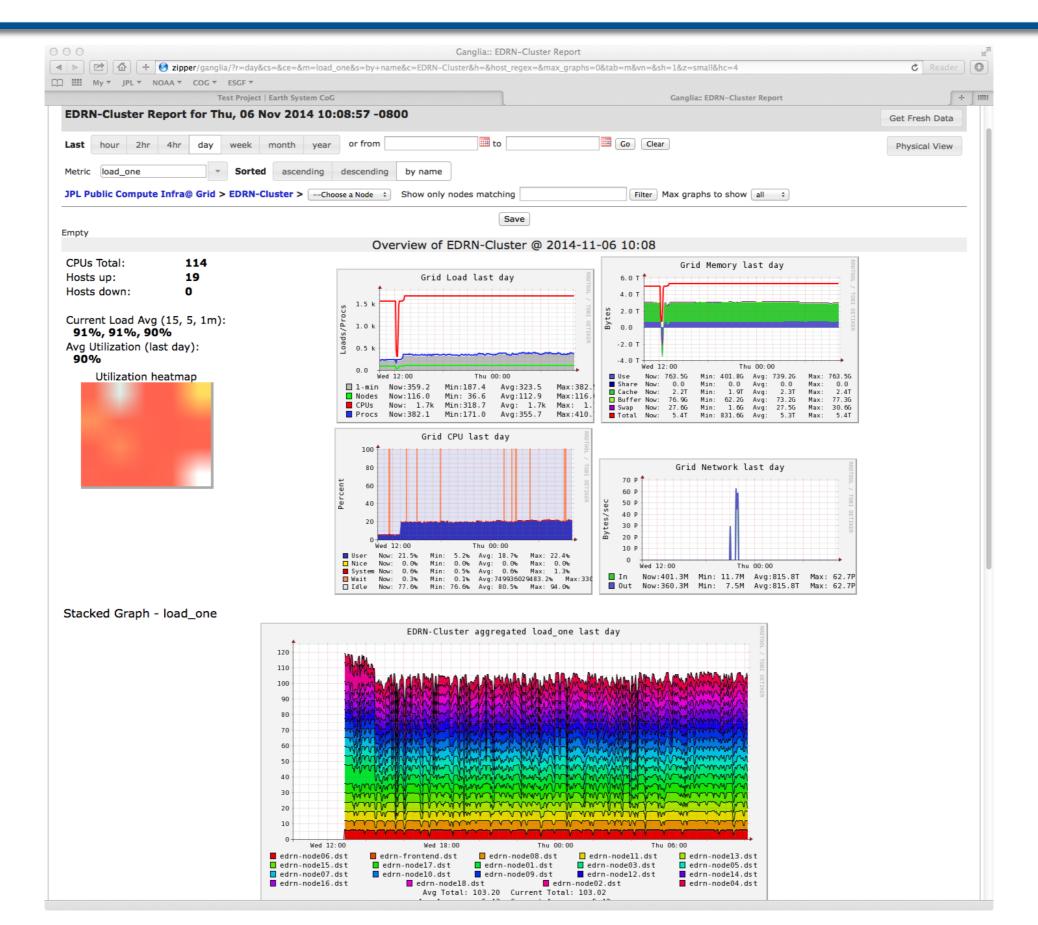






JPL/EDRN Cluster Usage (1 day)







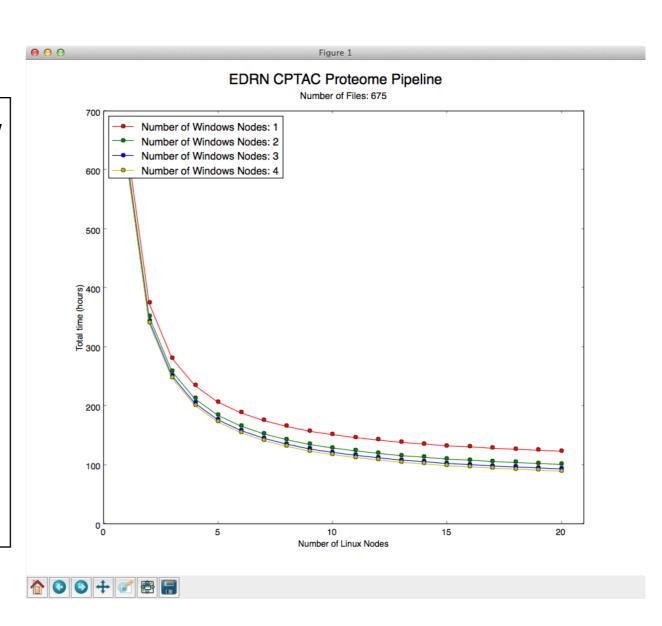
DAWN Simulation for Proteomics Pipeline



<u>DAWN</u> (<u>Distributed Analytics, Workflows and Numeric</u>): model to simulate and optimize Big Data computational pipelines, developed at JPL under the Data Science initiative

Analysis of Proteomics distributed workflow (using rough benchmarking):

- Clear reduction in overall elapsed time when MyriMatch step is executed on multiple nodes
- Efficiency gain levels off around 15 nodes
- Another gain is obtained by using 2
 Windows nodes instead of 1
- Full pipeline should complete in 4-5 days





Future Work



Proteomics Pipeline

- Re-run and benchmark full pipeline on new hardware for pancreatic test case
 - ▶ Submitted "report" to Vanderbilt to fix IdpPicker bug on Linux OS
- Run pipeline on other datasets
- Finalize pipeline product metadata
- Publish generated products to LabCAS

LabCAS

- Total redesign of Front-end Web Portal
 - ▶ Streamlined uploads of data files
 - ▶ Flexible UI for execution of data processing pipelines
 - ▶ Enhanced searching of datasets (by keywords or facets)
 - ▶ Improved access to documentation
- Back-end Services and Pipelines
 - ▶ Ability to package specific workflows for deployment at other institutions
 - ▶ Develop test suite
 - ▶ Upgrade to latest OODT release