

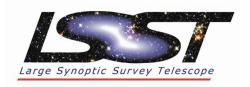
High-Volume Pipeline Processing with the LSST Data Management System

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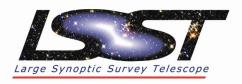
http://dev.lsstcorp.org

"Layers" of the LSST Data Management System (DMS)

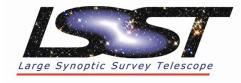


- Applications: libraries representing astronomically-relevant classes and algorithms implemented using those classes
- Middleware: libraries and tools that handle work-flow, I/O, information management (e.g.inputs provenance,) that are used to stitch algorithms together into processing pipelines.
- Infrastructure: Hardware architecture that runs the pipelines
- Software Environment: tools for building, installing, and using software

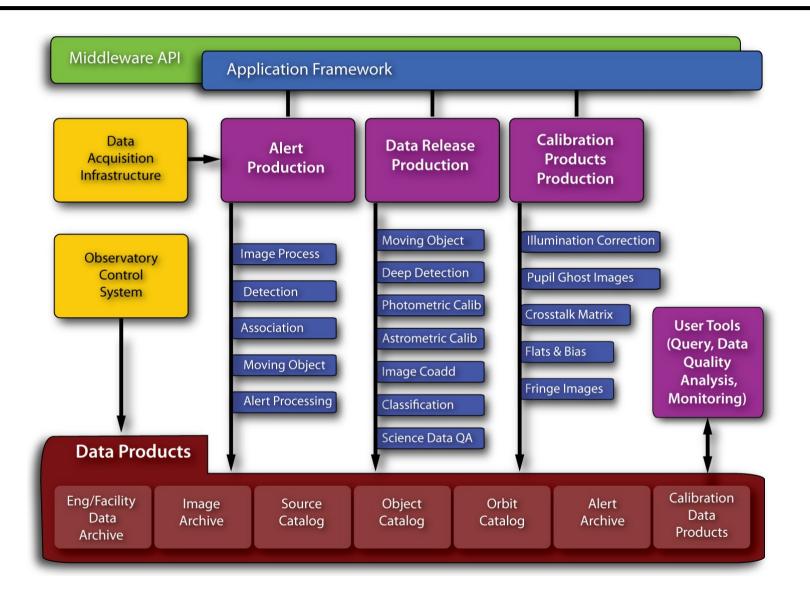
Software Environment



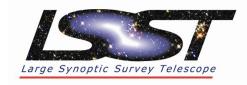
- DMS software stack: integrated set of 3rd-party and custom packages
 - 22 3rd-party packages, 25 custom package (distributed) + ~10 (in dev.)
- Custom packages (repos: http://dev.lsstcorp.org/trac/browser/DMS)
 - C++ for core reusable classes
 - SWIG used to create Python bindings
 - Python layer for connecting classes into applications and pipelines
 - Some Java and Jython
- Stack is highly modular
 - Allow one to minimize what gets installed
 - Requires that we track dependencies carefully
 - EUPS: a tool for managing the environments of a collection of interdependent packages
 - setup utils 3.0 updates the values of environment variables (PATH, LD_LIBRARY_PATH, PYTHONPATH, etc.) for package utils, version 3.0, and all its dependencies
 - Multiple versions can be installed at the same time
 - Provides rpm-like distribution of packages (built from source)



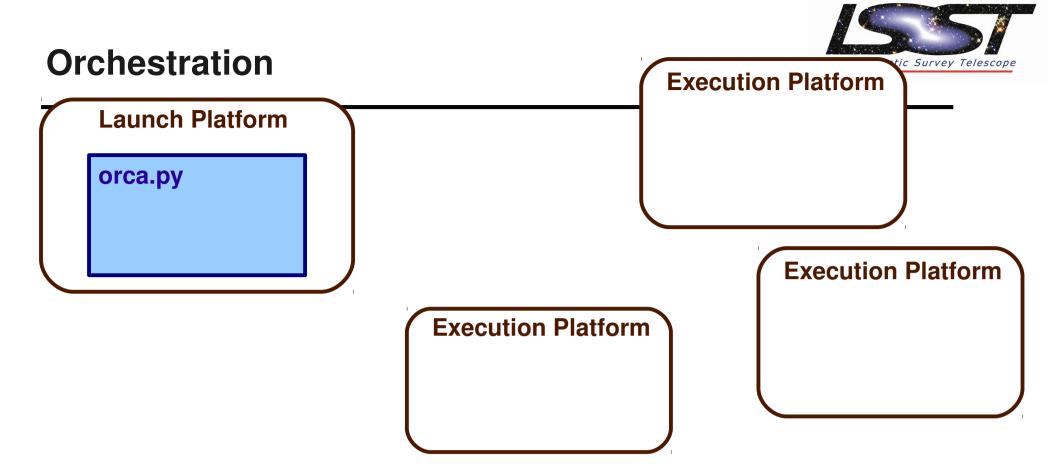
The Science View of LSST Processing





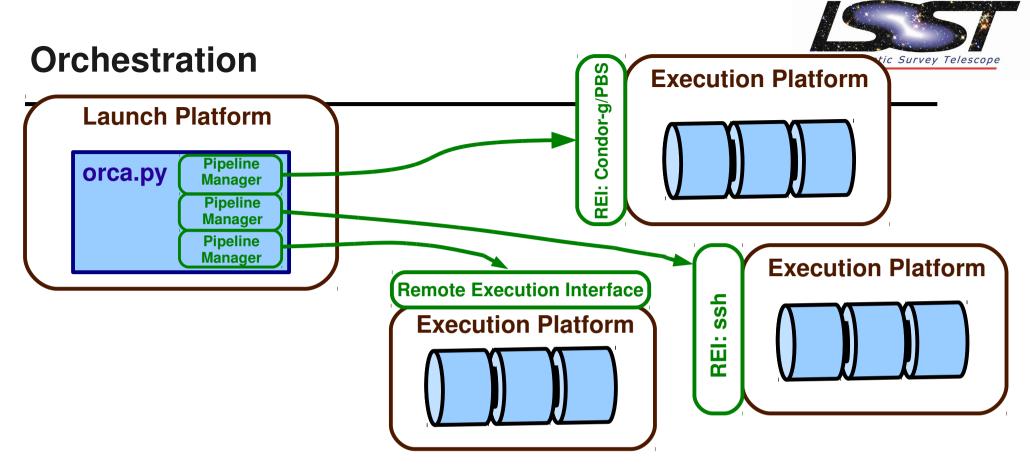


- Two categories of processing => two strategies
 - Alert Production: Real-time Processing
 - Executed nightly
 - Minimize I/O by keeping data in memory
 - Stress data-parallelism; requires consistent routing of data
 - Isolate and minimize parallel process cross-talk
 - Parallelism implemented using MPI
 - Data Release Production: High-volume Processing
 - Executed yearly but continuously
 - We can trade performance for robustness
 - Processing is more complex
 - with changing axes of parallelism
 - Parallelism implemented using Condor
- Categories include some common needs
 - Provenance tracking
 - Logging under high levels of parallelism
 - Encapsulated data access via logical identifiers



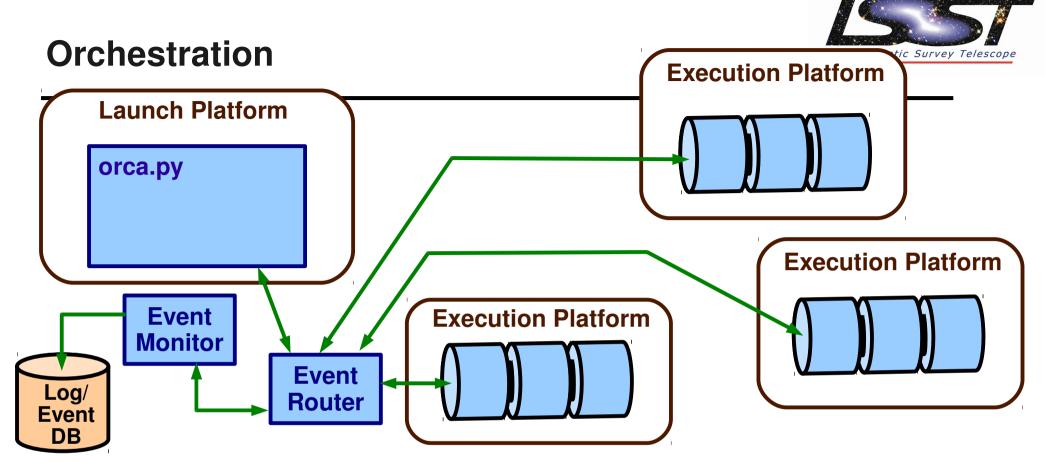
Some LSST vocabulary

- **Production** = a set of pipelines that collaborate to produce data products.
- Production run = an execution of the pipelines in a production
- Orchestration
 - The launching and monitoring of the pipelines for a production run
 - Applying the planning / strategies for adapting to a particular hardware platform and data being processed.



Orchestration layer launches pipelines on remote execution platforms

- adapts to different types of platforms and the way they run applications
- LSST designed to run on own dedicated platforms or community platforms (e.g. NCSA public resources)
- Launch mechanisms currently supported:
 - ssh
 - Condor-g: generic interface to local batch system (e.g. PBS)
 - Condor/DAGman



Pipelines communicate with each other and with the Orchestration layer via Events

- Event system based on the Java Messaging Framework (JMF)
- · Pipeline log messages sent out as events for remote recording
- An Event Monitor analyses progress to detect possible problems
 - Node failure, Runaway process, ...
 - Can signal orchestration layer to relaunch failed processing
- Inter-pipeline communication via Events
 - One pipeline may "wait" until an expected event with needed information arrives
 - Event payloads are light: one pipeline may tell another where to look for data.

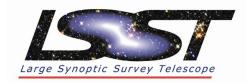




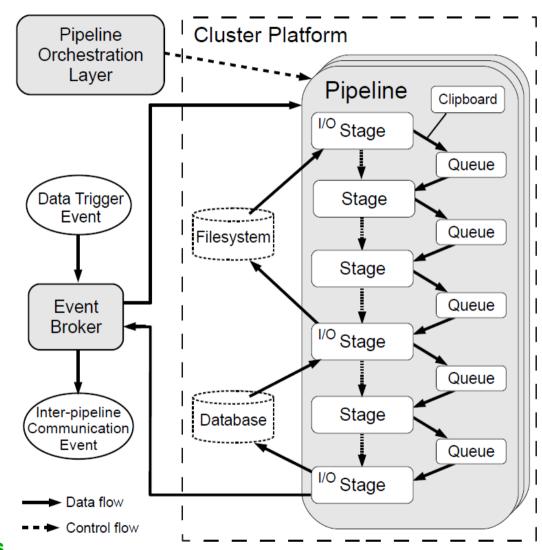
Three MPI-based pipelines

- Image Processing and Source Detection (IPSD)
 - Receives an event indicating which exposure to process
 - Removes instrumental signature, subtracts from template, detects source changes
 - Each MPI worker ("slice") operates on a different CCD segment
 - Tells Association Pipeline (via an event) where to find (in database) sources detected in difference image
 - Some "inter-slice" communication for cross-talk correction (not implemented) and WCS solution
- Moving Objects Prediction (MOPS)
 - Receives an event indicating a field of view on sky and a time
 - predicts positions of known moving objects
 - Each slice operates on a subset of objects
 - Tells Association Pipeline (via an event) where to find (in database) list of known moving objects in FOV
- Association Pipeline
 - Receives list of moving objects in FOV, list of difference sources via events
 - Matches difference sources with known sources





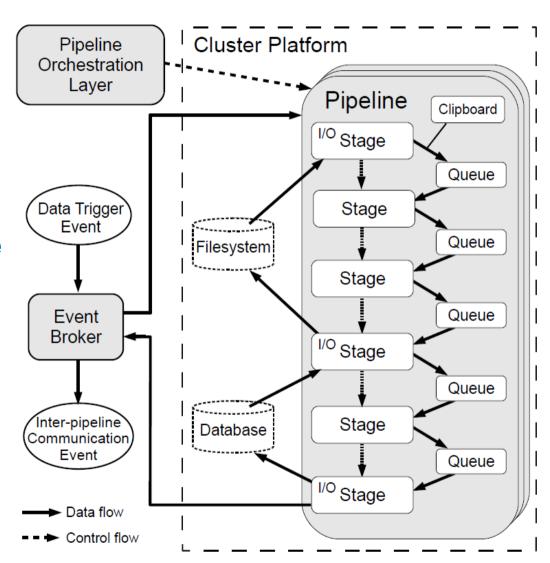
- Pipeline = a sequence of Stages
 - "Harness" = container for stitching together stages
- Pipeline has N+1 threads
 - Pipeline thread is the master controller
 - Slice threads process a dataparallel unit of data
 - e.g. CCD segment
 - All threads have same basic structure
- Sequence of Stages
 - Data queues sit b/w each stage
 - Data is passed from one stage to another through queues via a "Clipboard"
 - Clipboard = hierarchical dictionary
 - Stages look for input data on clipboard and place new data items on it.

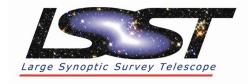




The Pipeline Harness

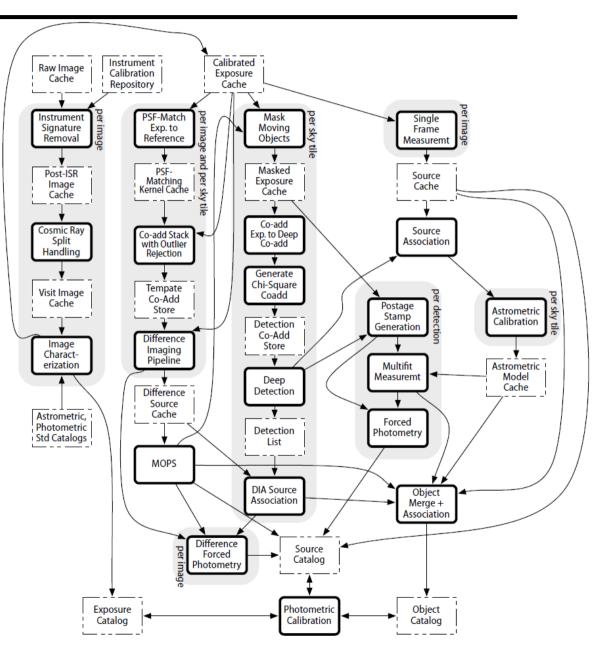
- I/O is done via dedicated I/O Stages
 - Algorithm stages are isolated from I/O
 - Output Stages can write products to disk at any time
- Processing is Event driven
 - A data trigger event signals the first stage which exposure to process next
 - Harness puts event data on clipboard
 - Each stage is run in sequence
 - After last stage is executed, pipeline returns to first stage, waiting for next event.



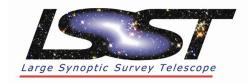


Data Release Pipeline (High-Volume)

- Sequencing is more complicated
 - Many interdependencies
 - Shifting "axes of parallelism"
 - By CCD segment
 - By sky-tile
 - By object
- Real-time is not required
 - Trade performance for robustness
 - Do more caching of data to disk
 - Leverage existing technology: Condor/DAGMan



Parallelism via Condor

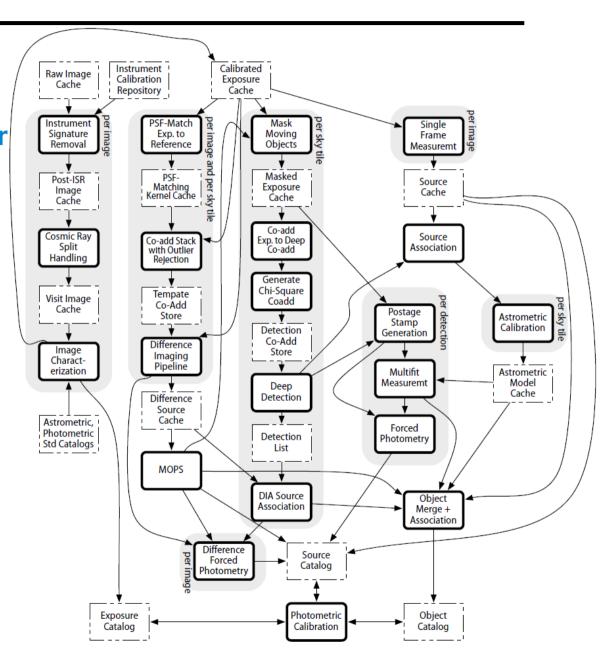


- Condor: parallel job execution system optimized for highly embarrassingly (data-) parallel applications
- DAGMan: a Make-like description of an application that can be processed by Condor.
 - Model app as jobs with interdependencies in form of a "directed acyclical graph" (DAG)
 - No inter-job communication
 - A job can be an MPI application to run on a cluster
 - Condor marches through DAG executing jobs while honoring dependencies
 - Typically, # of jobs >> # of available cores
 - Condor will maximize core utilization
- Can take advantage of Condor robustness features
 - Automatic rescheduling of failed jobs
 - Can detect loss of contact with jobs

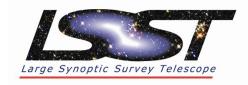
Large Synoptic Survey Telescope

Data Release Production via Condor/DAGMan

- Short and Narrow pipelines
 - Single slice, operating or single data-parallel unit of data
 - Only one unit will pass through pipeline per instantiation
- External process for staging input data
 - Volume to process >> capacity of disk
 - Will listen for events indicating progress of pipelines
 - Will stage data from mass storage before needed



Creating a Stage



- Harness provides a Stage API for turning a general algorithm into a pluggable pipeline component
 - Developer concentrates on applying algorithm to a data-parallel unit of data
 - No I/O included
 - No managing of parallel processing
 - Allows stage to be put into either Alert Production or Data Release
 - Stage has up to three steps:
 - Serial pre-processing (via Master Pipeline)
 - Parallel processing (via a Slice)
 - Serial post-processing (via Master Pipeline)
 - Note: Serial and Parallel processing on different nodes!
 - API allows passing data between parallel Slices and between master and slices (scatter/gather)
- Stages are configured via text configuration files call "policy files"
 - Hierarchical property dictionary
 - Simple, hand-editable file format
 - A special "Dictionary" format can be used to validate policy data
 - Much like XML Schema for XML documents

Building Community Software



- LSST Users will create advanced science data products
 - Will want to create new processing and analysis algorithms
 - Will want to tweak the application of standard processing
- LSST Users will want to leverage LSST DMS
 - Use framework to build advanced pipelines
 - LSST Data Access Centers will provide access to computing resources to create advance products
- An open, distributable software framework is, thus, an important requirement
 - Leverage other computing platforms
 - Allow application to other data and observatories