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Scaling cloud for LSST catalog at IN2P3



Fabrice Jammes

Scalable Data Systems Expert IN2P3

Frédéric Gaudet

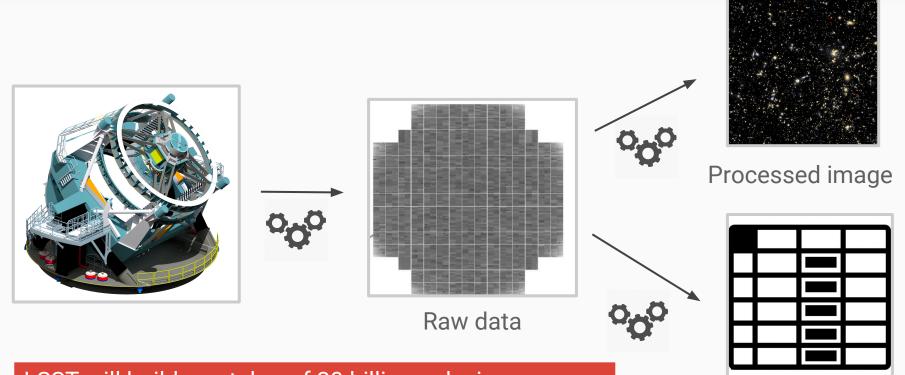
Openstack Gloud architect CNRS

Nicolas Chotard

Researcher

IN2P3

80+ PB of astronomical catalog



LSST will build a catalog of 20 billion galaxies and 17 billion stars and their associated physical properties

Catalog (stars, galaxies, objects, sources, transients, exposures, etc.)

Data

Images Persisted: ~38 PB Temporary: ~½ EB



★ ~3 million "visits"
★ ~47 billion"objects"
★ ~9 trillion "detections"

- ★ Largest table: ~5 PB
- ★ Tallest table: ~50 trillion rows
- ★ Total (all data releases, compressed):
 ~83 PB

Ad-hoc user-generated data Rich provenance

Database schema

1

Large Synoptic Survey Telescope LSST Database Schema Browser alpha Schema versions available for browsing: baseline DC3a PT1_1 PT1_2 ImSim 512_sdss 512_lsstsim (underlined showed) User defined functions documentation: version 0.1, version 0.2, version 0.3 (default on lsst10)						
Table List						
AAA_Version_3_2_4 ApertureBins CcdVisit CcdVisitMetadata DiaForcedSource	The Object table contains descriptions of the multi-epoch static astronomical objects, in particular their astrophysical properties as derived from analysis of the Sources that are associated with them. Note that fast moving objects are kept in the MovingObject tables. Note that less-frequently used columns are stored in a separate table called Object_Extra.					
DiaObject DiaObject_To_Object_Match	name	type	not null	unit	ucd	description
DiaSource ForcedSource	objectId	BIGINT	ý.		meta.id;src	Unique id.
LeapSeconds Object	parentObjectId	BIGINT				Id of the parent object this object has been deblended from, if any.
Object_APMean Object_Extra	procHistoryld	BIGINT	у			Pointer to ProcessingHistory table.
Object_NonPeriodic Object_Periodic	psRa	DOUBLE		deg	pos.eq.ra	RA-coordinate of the center of the object for the Point Source model at time 'psEpoch'.
prv_Amp	psRaSigma	FLOAT		deg	stat.error;pos.eq.ra	Uncertainty of psRa.
prv_Ccd prv_cnf_Amp	psDecl	DOUBLE	i i	deg	pos.eq.dec	Decl-coordinate of the center of the object for the Point Source model at time 'psEpoch'.
prv_cnf_Ccd prv_cnf_Filter	psDeclSigma	FLOAT		deg	stat.error;pos.eq.dec	Uncertainty of psDecl.
prv_cnf_Fpa prv_cnf_InputDataSet	psMuRa	FLOAT		mas/yr	pos.pm	Proper motion (ra) for the Point Source model.
prv_cnf_Node prv_cnf_Raft	psMuRaSigma	FLOAT		mas/yr	stat.error;pos.pm	Uncertainty of psMuRa.
prv_cnf_Run	psMuDecl	FLOAT		mas/yr	pos.pm	Proper motion (decl) for the Point Source model.
prv_cnf_Task prv_cnf_Task2TaskExecution	psMuDeclSigma	FLOAT	8 9	mas/yr	stat.error;pos.pm	Uncertainty of psMuDecl.
prv_cnf_Task2TaskGraph prv_cnf_TaskExecution	psParallax	FLOAT		mas	pos.parallax	Stellar parallax. for the Point Source model.
prv_cnf_TaskGraph prv_cnf_TaskGraph2Run	psParallaxSigma	FLOAT	í í	mas	stat.error;pos.parallax	Uncertainty of psParallax.
prv_Filter	uPsFlux	FLOAT		nmgy	phot.count	Calibrated flux for Point Source model for u filter.
prv_Fpa prv_InputDataSet	uPsFluxSigma	FLOAT		nmgy	stat.error;phot.count	Uncertainty of uPsFlux.
prv_Node prv_ProcHistory	gPsFlux	FLOAT		nmgy	phot.count	Calibrated flux for Point Source model for g filter.
prv_Raft prv_Run	gPsFluxSigma	FLOAT		nmgy	stat.error;phot.count	Uncertainty of gPsFlux.
prv_Snapshot	rPsFlux	FLOAT		nmgy	phot.count	Calibrated flux for Point Source model for r filter.
prv_Task		FLOUT	8 00			

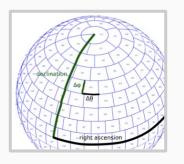
http://ls.st/s91

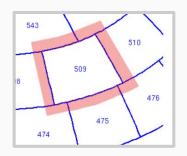
Implementation Strategy

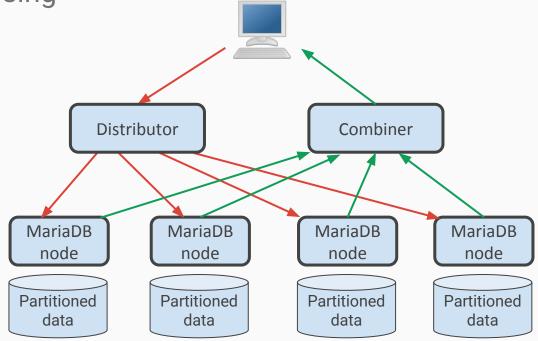
- ★ 100% Open source
- ★ Keep it flexible
- ★ Hide complexity
- \star Reuse existing components:
 - MariaDB, MySQL Proxy, XRootD, Google protobuf, Flask
- ★ Plus custom glue
 - C++, a bit of python, some ANTLR
 - Lots of multithreading, callbacks, mutexes and sockets
- ★ And custom UDFs

Qserv design

- ★ Relational database, spatially-sharded with overlaps
 ★ Map/reduce like processing
- ★ Map/reduce-like processing







On the french side

Who we are: French side

Research and Engineering

- ★ Oualid Achbal: Cloud-Computing, Cl
- Christian Arnault: Alternate solutions (MongoDB, Spark)
- ★ Sébastien Binet: Containers, Orchestration
- ★ 🛛 Nicolas Chotard: Data loading
- \star 🛛 Vincent Gatignol: Openstack, Cl
- ★ Frédéric Gaudet: Openstack, CEPH
- ★ Fabrice Jammes: Qserv development
- ★ Amine Mesmoudi: Alternate solutions (Hadoop)
- ★ 🛛 Bogdan Vulpescu: Data loading

















Who we are: French side, CC-IN2P3

Operation Team @ CC-IN2P3

- ★ Fabio Hernandez: Coordinator
- ★ Osman Aidel: Database, Spark
- ★ Yvan Calas: 50 nodes cluster management
- ★ Mathieu Puel: System Administration
- ★ Loïc Tortay: Shared Storage
- ★ Fabien Wernli: Monitoring





Fabio Hernandez











Fabien Wernli

Osman Aidel

Loïc Tortay

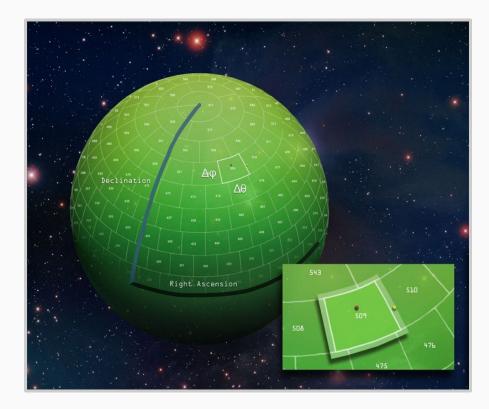
Mathieu Puel

Yvan Calas

What we do

Data Access and Database

- ★ Qserv scientific validation
- ★ Preparing production at Large Scale:
 - Continuous integration (SQUARE)
 - Integration and Large Scale tests
 - Deployment
 - Monitoring
 - Orchestration
- ★ Study and design deployment on modern infrastructure
 - Distributed storage: CEPH
 - Containerization
 - Kubernetes
 - Openstack
- ★ Alternate solution testing (Christian Arnault, A. Mesmoudi)

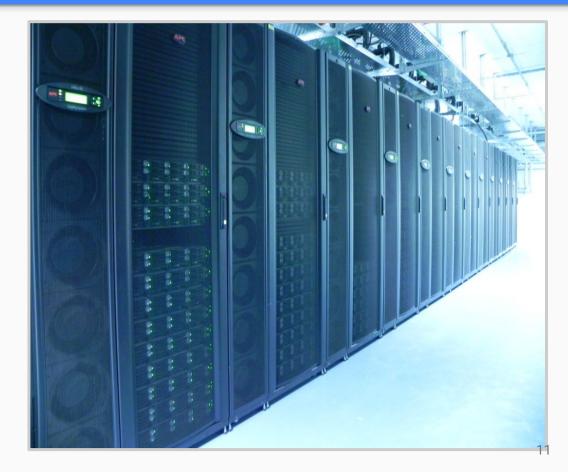


Tests and demonstrations

Target for production ~500 nodes clusters in 2 international data-centers

Running now Development platform (CC-IN2P3) 400 cores, 800 GB memory 500 TB storage, => ~65 TB data set on 2*25 nodes

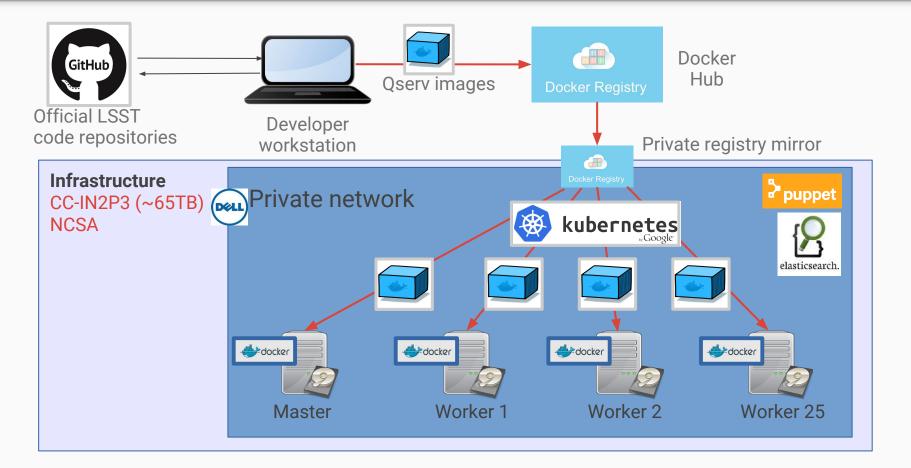
Prototype Data Access Center (NCSA) 500 cores, 4 TB memory 700 TB storage, WISE data loaded



Large Scale Tests: All about the data

- ★ The data set comes from the Stripe 82 released by the SDSS telescope. It has been processed by the Qserv team and duplicated to reach 35 TB in 2015 and 65 TB in 2017.
- ★ No real scientific meaning (yet !)

Automated deployment: bare-metal



Petasky

- ★ PetaSky : « Gestion et exploration des grandes masses de données scientifiques issues d'observations astronomiques grand champ »
- ★ PetaSky uses test dataset from both LSST and Euclid
- ★ Involved labs : LIMOS, LIRIS, LPC, APC, LAL, LaBRI, LIF, LIRMM, LAM, CC-IN2P3

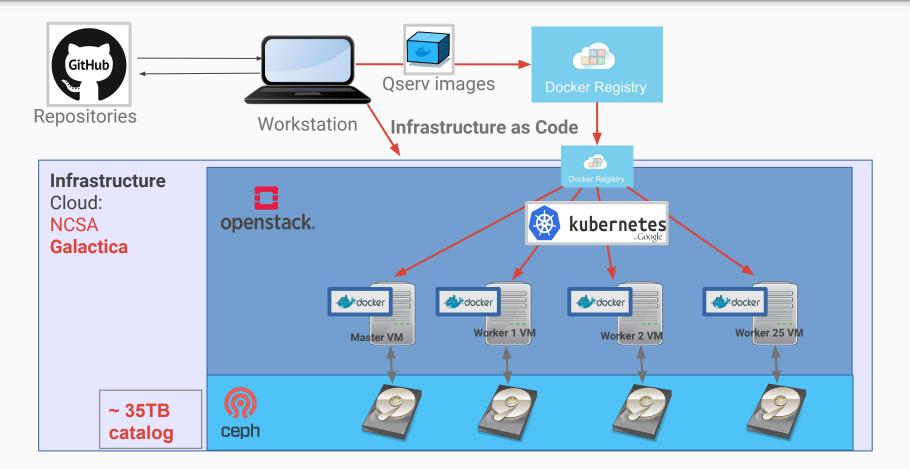
Galactica: experiment objectives

- ★ Set up a Large Scale Continuous Integration platform: benchmark QServ releases against a 35TB data set
- ★ Prototype Qserv deployment/orchestration over the cloud QServ will land in production datacenter (CC-IN2P3/NCSA)
- ★ Experiment how Cloud can scale to Big-Data OpenStack compute nodes + Ceph storage

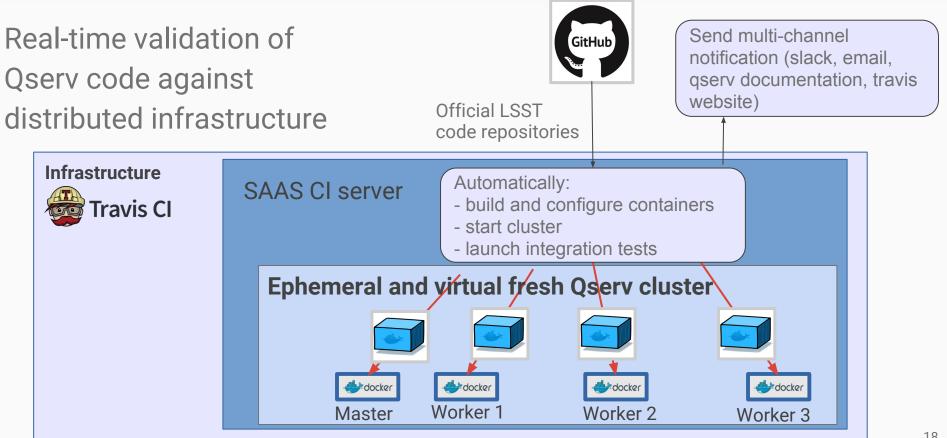


- ★ CEPH side: 24*1,5 TB virtual disks —> 35TB of data
- ★ Openstack: 25 VM using 16GB RAM and 2 vCPU each

Automated deployment: Openstack+CEPH



Automated deployment: advanced CI



What does bring an elastic infrastructure?

Openstack+containers In a few seconds:

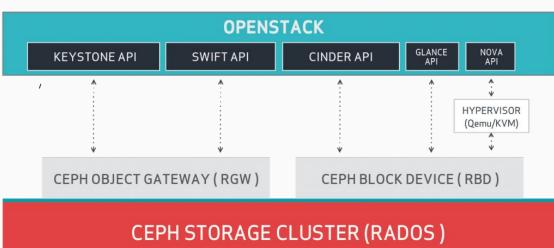
- Provision a Qserv cluster
- Package and deploy development version

CEPH

Enable large storage access via Openstack Provide:

- Data replication
- Data high availability
- Data reconstruction

=> Currently trying to make it scale



What does bring an elastic infrastructure?

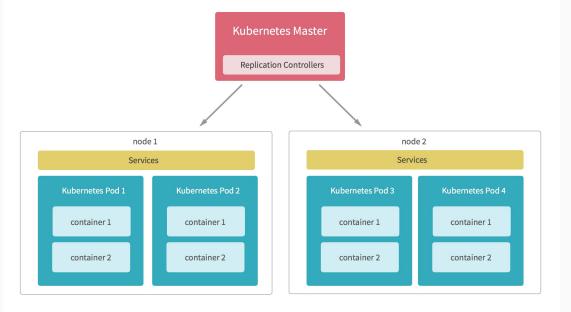
Kubernetes

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Orchestrate Qserv processes

Can scale to hundred of nodes **Provide**:

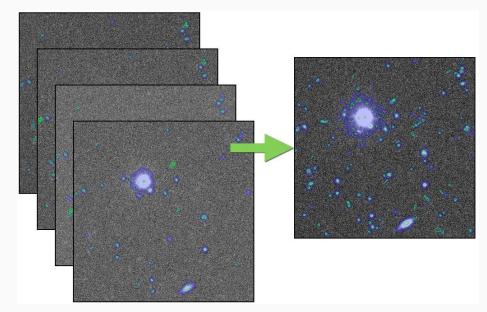
- Container placement
- Auto-scaling
- Auto-healing
- Volume management (storage)
- , Resource usage monitoring
- Health checks
- Rolling update



Qserv integration into science pipelines

Test Qserv on real data processed through the LSST stack

- ★ Process a dataset, and produce the catalogs using LSST Stack
- ★ Create a Qserv instance ready to ingest data
- ★ Transform the catalogs into a data format that Qserv can ingest
- \star Load the catalogs into Qserv
- ★ Create a set of queries to test basic Qserv features
- \star \prime Run these queries using python
- ★ Implement into Clusters pipeline
- \star Extend to other analysis



Conceptual data model for stack output ?

Alternate solutions

Hive/HadoopDB (A. Mesmoudi & al)

Benchmarking SQL on MapReduce systems using large astronomy databases

DOI: 10.1007/s10619-014-7172-8

MongoDB (C. Arnault)

Test using Qserv S15 dataset

- Data for one worker node (1.3 TB)
- Using Openstack/Galactica

Very promising results on simple queries (i.e. selection on indexed fields)

Investigating on:

- ★ Joins between large tables
- ★ Near-neighbors
- ★ Non-indexed selections

★ Data-distribution (chunk overlap distribution) Moving toward Spark/Dataframe/GeoSpark









XLDB 2017 in Europe

Session and chairs:

Polystores

Patrick Valduriez, INRIA: Senior researcher, head of Zenith research team.

Applications: earth and astronomy, neuroscience

- Peter Baumann, Jacobs University: Professor and head of the Large-Scale Scientific Information Systems research group.
- Romulo Goncalves. Nederland eScience Center: Expert in • Databases, Data Structures, Distributed Computing.

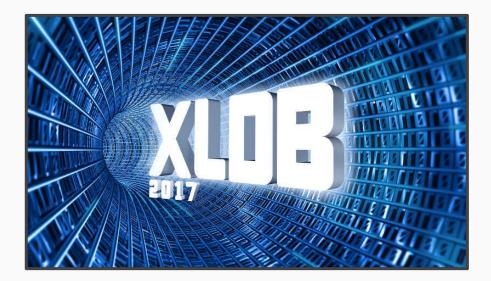
Modern data management

- Anastasia Ailamaki. EPFL: Professor and Lab Director.
- Mohand-Said Hacid. LIRIS: Professor and Lab Director.

Scaling Cloud to Big Data

Dirk Duellmann, CERN: Deputy leader of the data and storage • services group in CERN's IT.

1.5 days for main conference ~1 day for Hackaton





\$ databricks





See you in autumn 2017 in Clermont-Ferrand http://xldb2017.uca.fr



Summary

★ French scientists are very interested in LSST catalog challenge

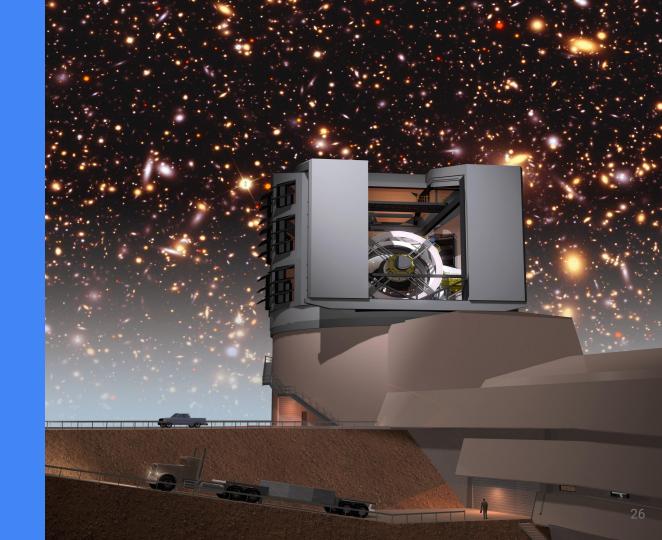
- Alternate solution testing
- Elastic deployment, prepare production
- Integration with LSST stack
- ★ Want to learn more?
 - <u>http://ls.st/4gh</u> (Database Design doc)
 - <u>http://ls.st/6ym</u> (User Manual)
- Are you an adventurous super early adopter? You can try it now
 - <u>http://ls.st/89y</u> (Qserv Documentation)

Thanks!

Contact:

Fabrice JAMMES LPC Clermont-Ferrand

fabrice.jammes@in2p3.fr



Implementation Details

