

Cloud Technologies in High-Energy Physics Data Processing

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Rob Joyce, NSA, Head of Tailored Access Operations

Cloud computing is a really fancy term for someone else's computer.

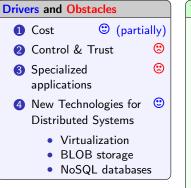


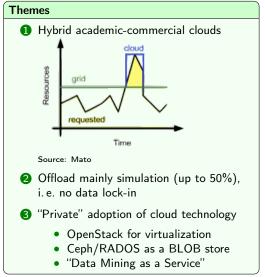
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Cloud computing is a really fancy term for someone else's computer.

- High-energy physic's idea of cloud computing is mostly limited to "Infrastructure-as-a-Service" (IaaS)
- 2 I will focus in this talk at cloud computing in the context of a means to improve our distributed systems ("the grid")









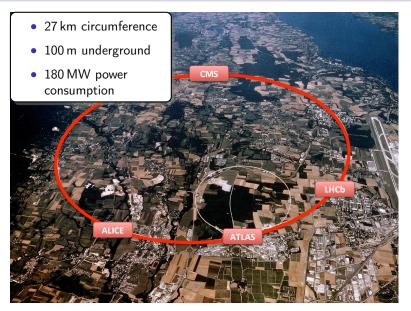


- 1 HEP Computing Model
- 2 Node Virtualization
- **3** Scientific Application Delivery
- 4 Volunteer Cloud
- 6 Cloud Storage
- 6 Summary & Outlook

HEP Computing Model

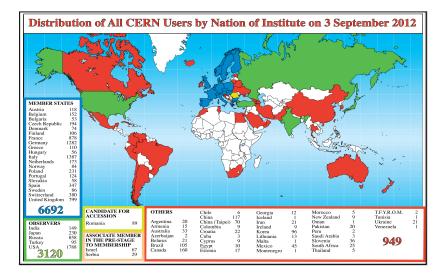


Bird's Eye View of the Large Hadron Collider



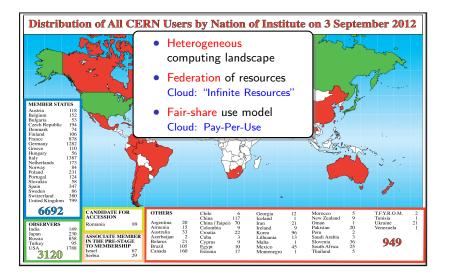


CERN Members and Users



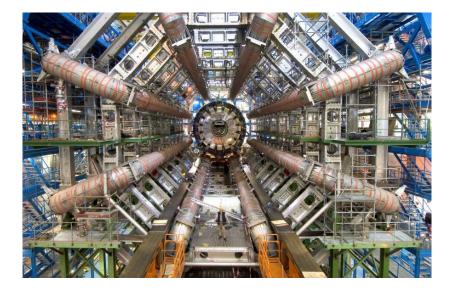


CERN Members and Users





The ATLAS Detector



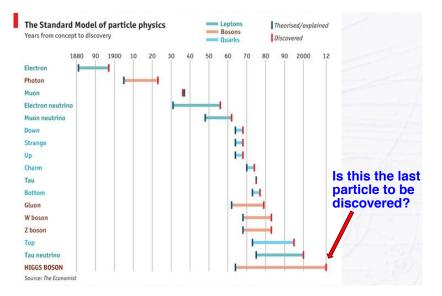


The ATLAS Detector

- 100 Million channels, bunch crossing every 25 ns
- 1 PB/s internal data rate
- 5 PB data / year recorded (without simulation) HL-LHC: increase to 100 PB / year by 2025 (×20)
- 5 M lines of experiment-specific code

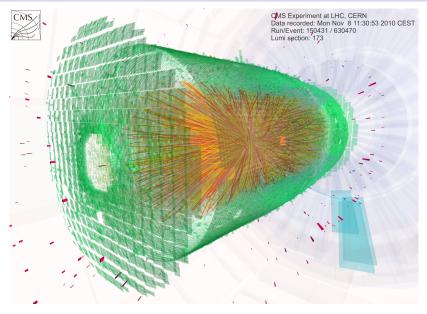


Time Line of the Standard Model

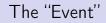




The "Event"







• Raw Event

Megabytes or tens of megabytes per event Sensor signals corresponding to a single crossing of beams

• Event Summary Data (ESD)

Hundreds of kilobytes per event

Reconstructed the physics information from digitized sensor signals. For instance: transformation in global time and space coordinates, reconstruction of trajectories, determination of particles' charge and momentum

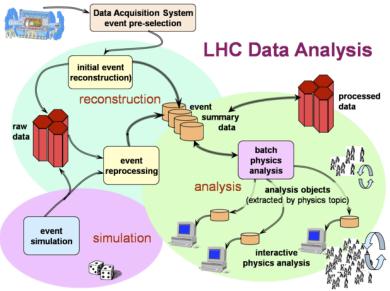
Analysis Object Data (AOD)

Tens of kilobytes per event Stripped ESD tailored to specific analysis goals

> Events are independent from each other, embarrassingly parallel workload

LHC Data Flow

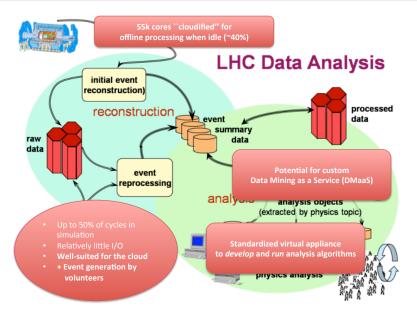




Source: Harvey et al.



LHC Data Flow



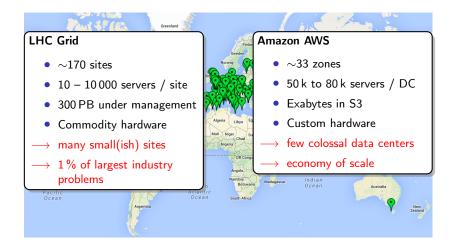


Distributed Processing of LHC Event Data



Source: WLCG

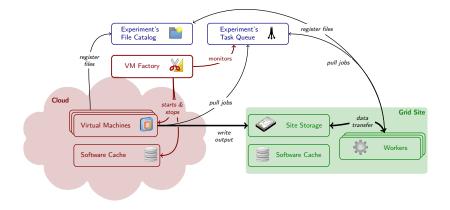




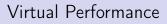
Node Virtualization



Cloud Resources in a Global Batch System









- Good VM performance requires careful configuration of hypervisor and guest
- High-performance I/O devices (e.g. 10 Gbit NIC, GPGPU, Xeon Phi, ...) only usable with direct assignment to VMs
- Just beginning to explore non x86_64 architectures (ARM64, POWER8)

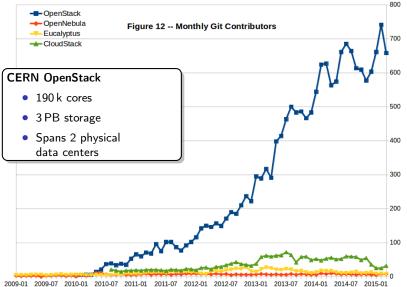
Performance engineering on CERN OpenStack:

| VM sizes (cores) | Before | After |
|---------------------|--------|------------------------------|
| 4x 8 | 7.8% | 3.3% (batch WN) |
| 2x 16 | 16% | 4.6% (batch WN) |
| 1x 24 | 20% | 5.0% (batch WN) |
| 1x 32 | 20.4% | 3-6% (bare SLC6 batch WN) |
| ▶ HEPiX'15 | Sou | rce: Wiebalck et al. |

Containers (e.g. Docker) provide isolated namespaces on top of the same kernel and have almost native performance.



Open Source Cloud Software



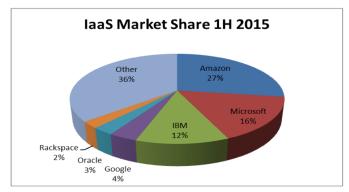
^{2009-01 2009-07 2010-01 2010-07 2011-01 2011-07 2012-01 2012-07 2013-01 2013-07 2014-01 2014-07 2015-01 2009-04 2009-10 2010-04 2010-10 2011-04 2011-10 2012-04 2012-10 2013-04 2013-10 2014-04 2014-10 2015-03}



Commercial Clouds

Market Share

• Amazon (EC2), Microsoft (Azure), IBM (OpenStack,...), Google Cloud Platform, Rackspace (OpenStack/Azure/VMware), ...



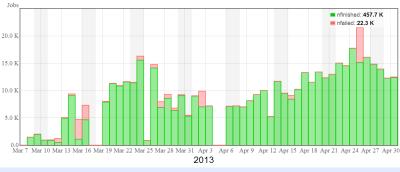
Ralph Finos, http://wikibon.com/public-cloud-market-shares-2014-and-2015/

Source: McNab



Commercial Clouds ATLAS @ Google Cloud Platform 2013

Failed and Finished Jobs



Most of the job failures occurred during start up and scale up phase – as expected
 Reached throughput of 15k jobs per day

Source: Panitkin

Google I/O

CERN

Scale Testing of Commercial Clouds

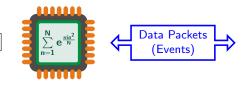
HEPiX'15 Brookhaven National Lab / Amazon EC2 100 000 cores Amazon spot market biddings HNSciCloud **CERN** / Evaluation of Various Commercial Clouds HelixNebula Science Cloud. EU funded project for federated clouds • Current tests at a few thousand cores but ramping up ~5k cores allocated on MS Azure A few thousand cores on Deutsche Börse Cloud Exchange Lessons Learned Cloud resources require supporting services besides the VMs (caches, proxies) We need to figure out our cost of computing Cloud resources need to be continuously benchmarked More layers of complexity that can cause operational issues

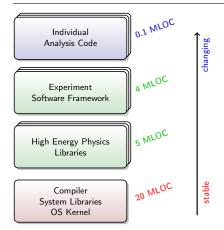
Application Delivery



High-Energy Physics Software Stacks

> cmsRun DiPhoton_Analysis_cfg.py



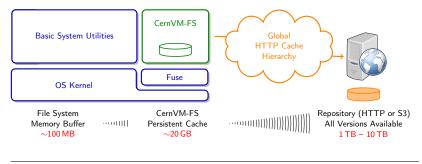


Scientific Software is Special

- Frequent Updates
- Includes its dependencies
- Released versions need to stay available
- Quick turn-around desired from source code commit to deployment



Application Delivery with the CernVM File System

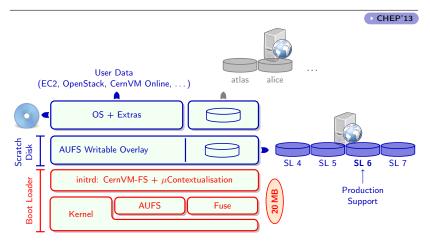




- Independent repositories, e.g. /cvmfs/atlas.cern.ch
- Single point of publishing
- HTTP Transport, access and caching on demand
- > 100 Million files, \sim 100,000 deployed clients (Grid, Cloud, HPC)



Twofold system: μ CernVM boot loader + OS delivered by CernVM-FS



 \sim 30 000 CernVM booted per day



The success of CernVM is largely based on the fact that it runs in practically $% \left({{\mathbf{H}}_{\mathbf{M}}} \right)$ all cloud environments.

| Hypervisor / Cloud Controller | Status |
|-------------------------------|-----------------------|
| VirtualBox | \checkmark |
| VMware | \checkmark |
| KVM | \checkmark |
| Xen | \checkmark |
| Microsoft Hyper-V | \checkmark |
| Vagrant | \checkmark |
| OpenStack | \checkmark |
| OpenNebula | \checkmark |
| CloudStack | \checkmark |
| Amazon EC2 | \checkmark |
| Google Compute Engine | \checkmark |
| Microsoft Azure | \checkmark |
| Docker | √ ¹ |







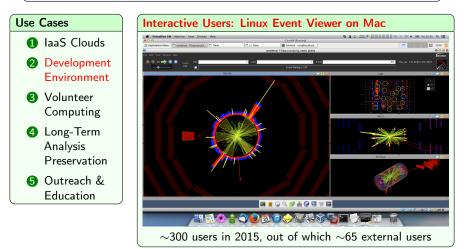
Infrastructure-as-a-Service Cloud

Various clouds:

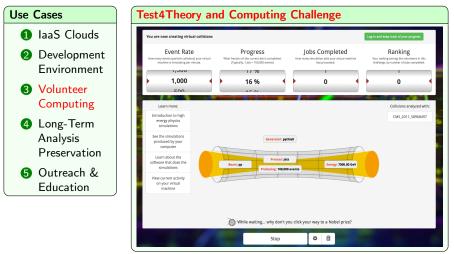
- ATLAS high-level trigger farm
- Cloud resources seamlessly integrated with experiment task queues (e.g. ATLAS CloudScheduler, LHCb VAC)
- Commercial providers
- ALICE software release testing on CERN OpenStack

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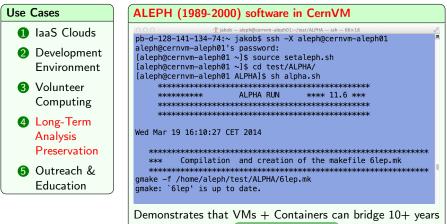






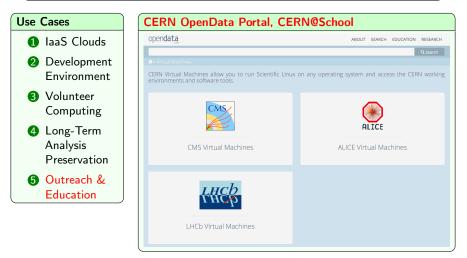






DPHEP Status Report





Volunteer Computing



X@Home

- Commonly known as@Home projects
- Computing on *spare* resources of "interested citizens"
- Opportunistic, volatile resources
- Big projects comparable to a TOP500 supercomputer

SETI@Home

- Search for extraterrestrial life
- 130 000 active participants
- 630 Tera-FLOPS



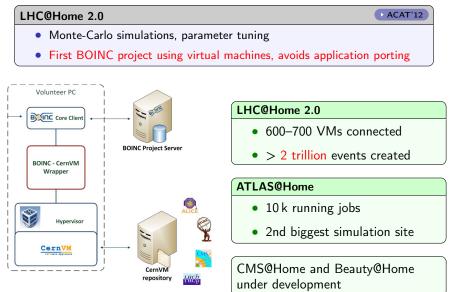
- Outreach program
- Requires curation of volunteers

Einstein@Home

- Search for gravitational waves
- 34 000 active participants
- 470 Tera-FLOPS



CERN

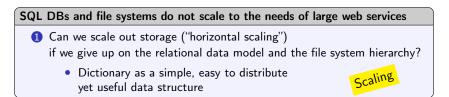


Source: Harutyunyan

Cloud Storage

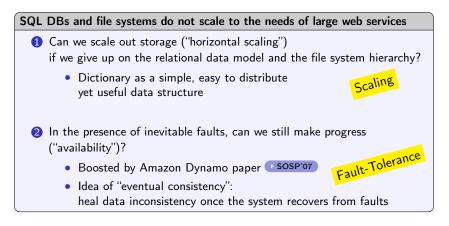


 \sim 2000–2010



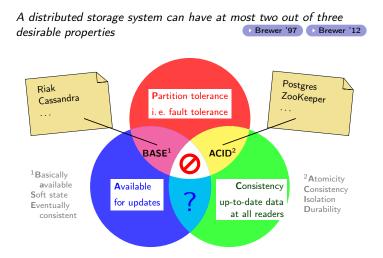


 \sim 2000–2010



Even though the *interface* is simple, the *implementation* of a distributed key-value or BLOB store is highly non-trivial.





The tradeoffs between availability and consistency can be granular and subtle For instance: a disconnected ATM might still allow small withdrawals

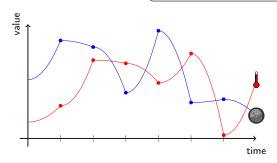


- CMS evaluated Cassandra, MongoDB, Riak
 CHEP'15
- ALICE evaluated RAMCloud and Riak

Note

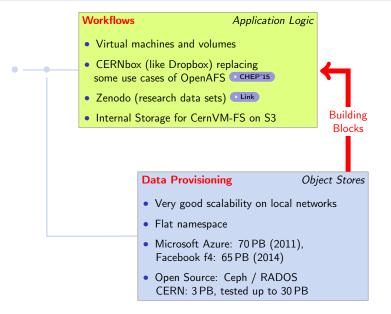
Conditions Data

- State of the detector at the time of data taking
- Small (terabytes / year) but critical to reconstruct data
- Traditionally kept in SQL databases



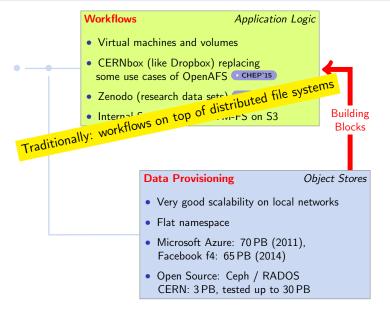


Workflows on Top of BLOB Storage





Workflows on Top of BLOB Storage



Summary & Outlook





Source: Piparo



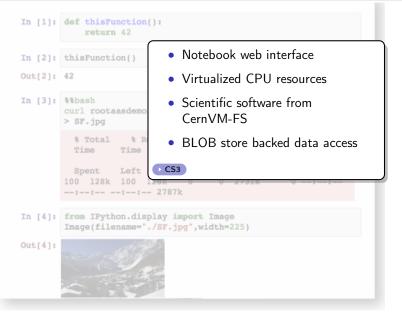
The Future: Custom Cloud Apps?

| In a brov | vser notion(): |
|-----------|---|
| In [2]: | thisFunction() |
| Out[2]: | 42 |
| Co | deaasdemo.web.cern.ch/rootaasdemo/SaasFee.jpg \ > SF.jpg |
| | % Total % Received % Xferd Average Speed Time Time Time Current |
| | Spent Left Speed 100 128k 100 128k :-: |
| In [4]: | <pre>from IPython.display import Image Image(filename="./SF.jpg",width=225)</pre> |
| Out[4]: | Images |

Source: Piparo



The Future: Custom Cloud Apps?





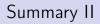
Cloud Computing Resources in HEP

- \sim 5 % of resources provided by virtual machines, mostly for simulation
- Cost of commercial VMs almost on par with in-house cost
- Volunteer clouds: great opportunity for outreach

Cloud Storage Resources in HEP

- Storage: more expensive commercially because of data transfer costs
- Scalable key-value interface requires more application logic
- Very promising open source components: Ceph, Riak, RAMCloud, ...





Did utility computing arrive?

 On the infrastructure level: x86_64 CPUs and unstructured storage are commoditiy

2 HEP still has unique and specialised needs and applications:

- Wide-area data management
- Custom, optimized I/O format
- Complex scientific software stacks
- Legacy software
- Our custom distributed systems must support our custom application workflows
- We have today much better building blocks available than at the time when we designed the grid

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