PanDA: распределенная система обработки Больших Данных в неоднородной компьютерной среде

Пятый Московский Суперкомпьютерный Форум

Алексей Климентов Владимир Кореньков Василий Велихов

21 Октября 2014









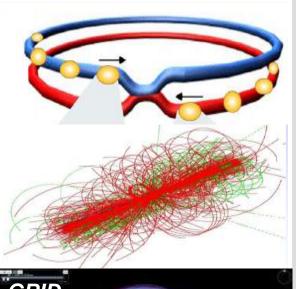
Main topics

- CERN and Large Hadron Collider (LHC)
 - LHC data processing challenges
 - Evolution of LHC computing model
 - Workload Management System core ideas.
- Workload Management System PanDA
- LHC Computing needs for Run2+ (2015-2018)
- megaPanDA project :
 - Expanding PanDA beyond the Grid and High Energy Physics (HEP)
 - PanDA at Super-Computing Facilities
- Summary

Tools: LHC and Detectors



Proton-Proton Collisions at the LHC



LHC delivered billions of collision events to the experiments from proton-proton and proton-lead collisions in the Run1 period (2009-2013)

- → collisions every 50 ns= 20 MHz crossing rate
- 1.6 x 10^{11} protons per bunch at $L_{pk} \sim 0.8 x 10^{34}$ cm²/s ≈ 35 pp interactions per crossing pile-up
- $\rightarrow \approx 10^9$ pp interactions per second !!!
- in each collision≈ 1600 charged particles produced

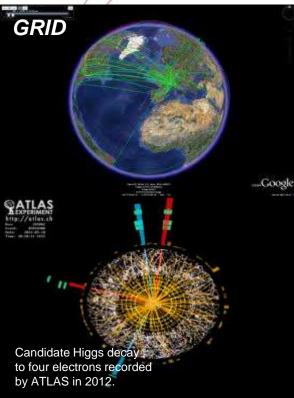
enormous challenge for the detectors and for data collection/storage/analysis

Raw data rate from LHC detector: 1PB/s

This translates to Petabytes of data recorded world-wide (Grid)

Up to 6 GB/s to be stored and analysed after filtering

The challenge how to process and analyze the data and produce timely physics results was substantial, but at the end resulted in a great success

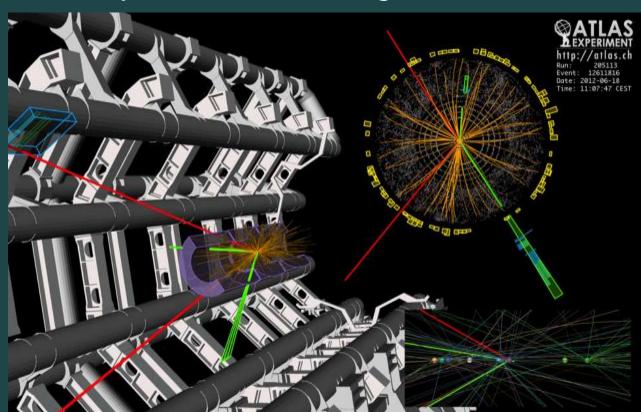


What is this data?

Raw data:

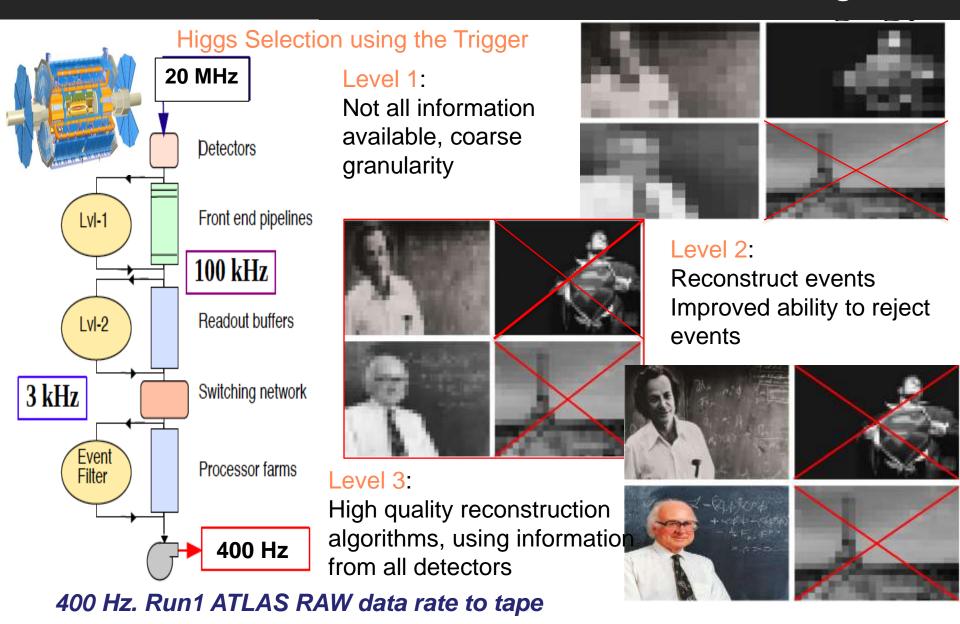
- Was a detector element hit?
- How much energy?
- What time?
- Reconstructed data:
 - Momentum of tracks (4-vectors)
 - Origin
 - Energy in clusters (jets)
 - Particle type
 - Calibration information

- 150 Million sensors deliver data ... 40 Million times per second
- Up to 6 GB/s to be stored and analysed after filtering



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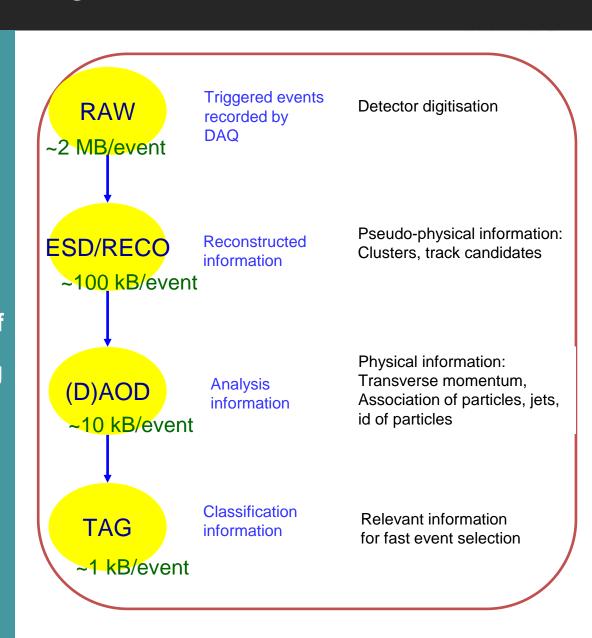
HEP Online. Reduce the data volume in stages.



_____6

HEP Offline. Data and Algorithms.

- HEP data are organized as Events (particle collisions)
- Simulation,
 Reconstruction and
 Analysis programs
 process "one event at a time"
 - Events are fairly independent → Trivial parallel processing
- Event processing
 programs are composed of
 a number of Algorithms
 selecting and transforming
 "raw" event data into
 "processed"
 (reconstructed) event data
 and statistics
- ~4 million lines of code (reconstruction and simulation)
 - ~1000 software developers on ATLAS





BNL/PAS

Some history of the scale...

Date	HEP Collaboration sizes	Data volume, archive technology
Late 1950's	2-3	Kilobits, notebooks
1960's	10-15	kB, punchcards
1970's	~35	MB, tape
1980's	~100	GB, tape, disk
1990's	700-800	TB, tape, disk : LEP
2010's	~3000	PB, tape, disk : LHC

For comparison:

1990's: Total LEP data set ~few TB Would fit on 1 tape today

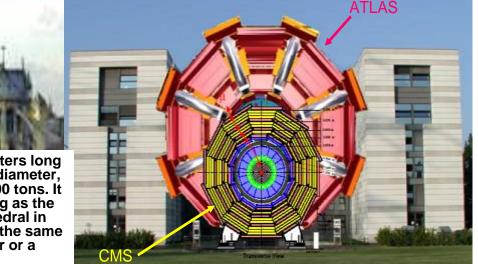
Today: 1 year of LHC data ~25 PB



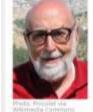
The ATLAS Experiment at the LHC



174 Universities and Labs From 38 countries More than 1200 students



The Nobel Prize in Physics 2013. François Englert, Peter Higgs The Nobel Prize in Physics 2013





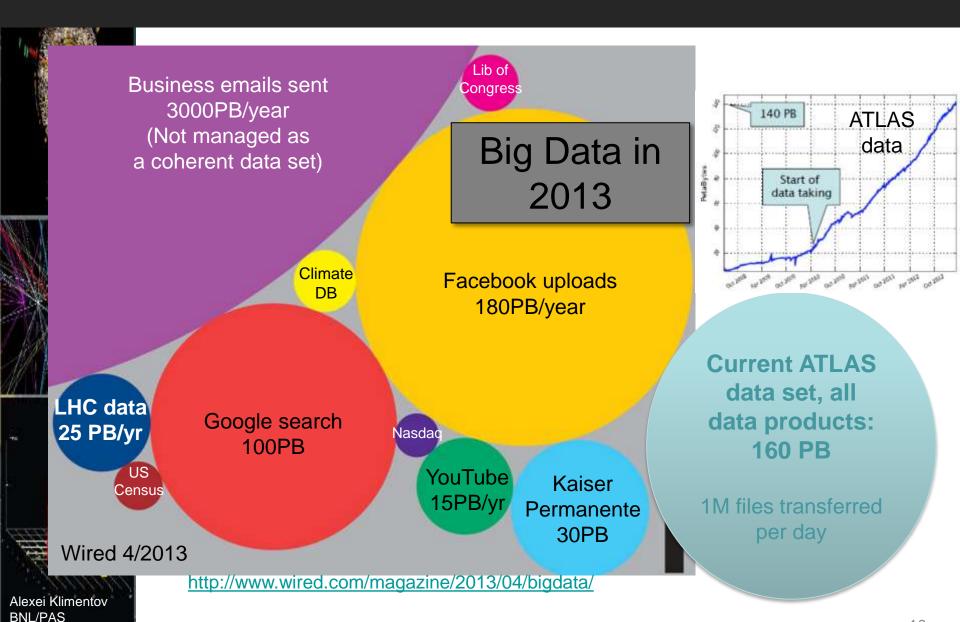


The Nobel Prize in Physics 2013 was awarded jointly to François Englert and Peter W. Higgs. "for the theoretical discovery of a mechanism that contributes to our understanding of the origin of mass of substamic particles, and which recently was confirmed through the discovery of the predicted fundamental particle, by the ATLAS and CMS experiments at CERN's Large Hadron Collider*

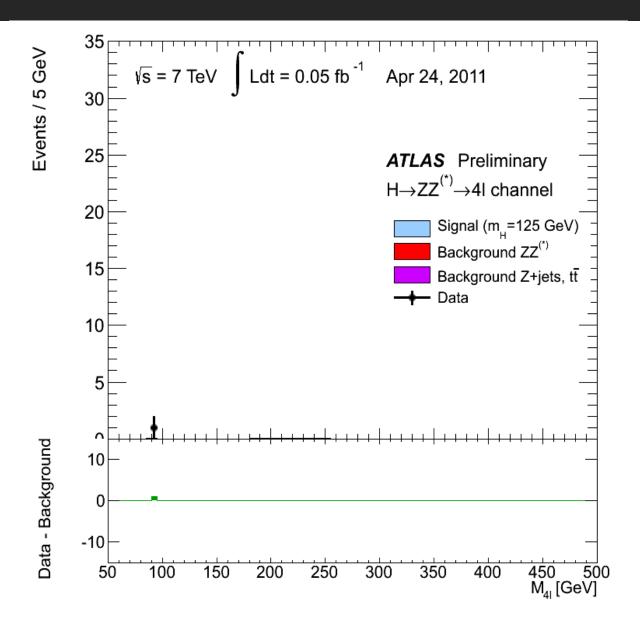
ATLAS has 44 meters long and 25 meters in diameter, weighs about 7,000 tons. It is about half as big as the Notre Dame Cathedral in Paris and weighs the same as the Eiffel Tower or a

hundred 747 jets

Big Data: often just a buzz word, but not when it comes to ATLAS...

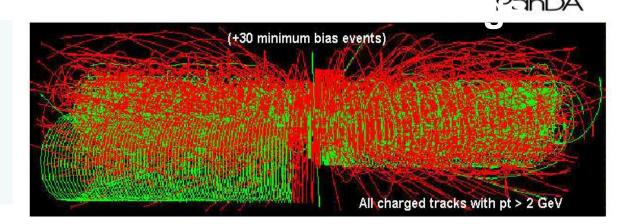


Higgs Boson Discovery



ATLAS Data Challenge

- 800,000,000 protonproton interactions per second
- 0.0002 Higgs per second
- ~150,000,000 electronic channels

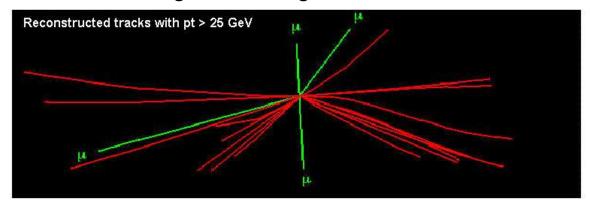


Selectivity: 1 in 10¹³

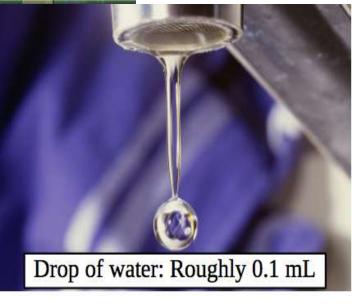
Like looking for 1 person in a thousand world populations

Or for a needle in 20 million haystacks!









New physics rate ~ 0.00001 Hz

Event Selection:

1 in 10,000,000,000,000

Like looking for a single drop of water from the Geneve Jet d'Eau over 2+ days





Offline Computing in High Energy Physics



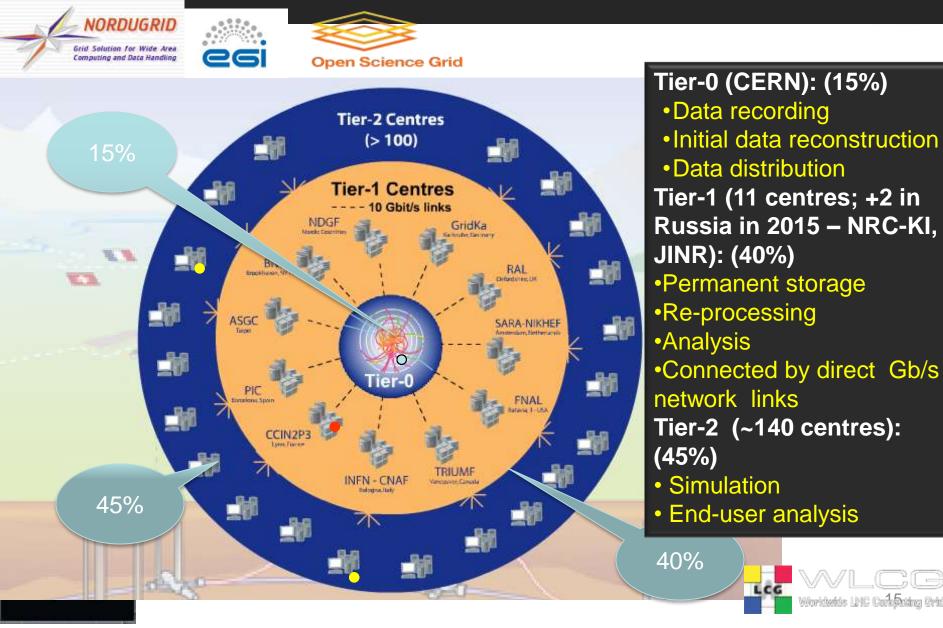
Offline computing in HEP

- Has changed and evolved dramatically over the past decades
- Especially for the biggest experiments at the LHC

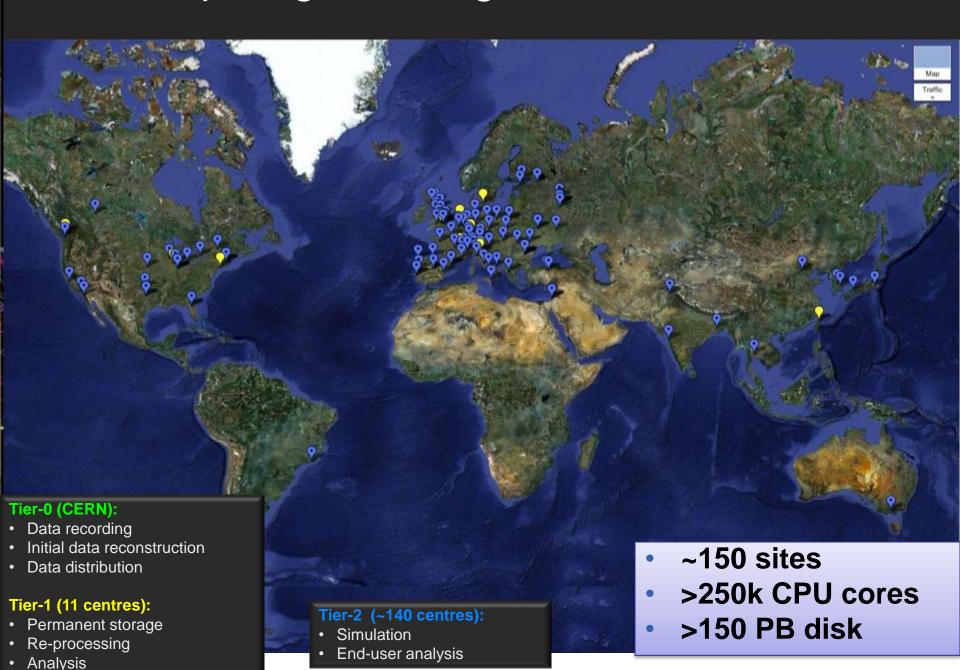
The situation ~15 years ago

- Data processing was performed in large computing centers using local batch systems with dedicated shares
- A few satellite centers did simulations, occasionally data reprocessing
- Users were mostly located near large computing centers, usually at the laboratory where the experiment was located, and used a combination of desktops and batch systems for analysis
- Final Data Summary Files versions were physically shipped for remote analysis

World-wide LHC Computing Grid



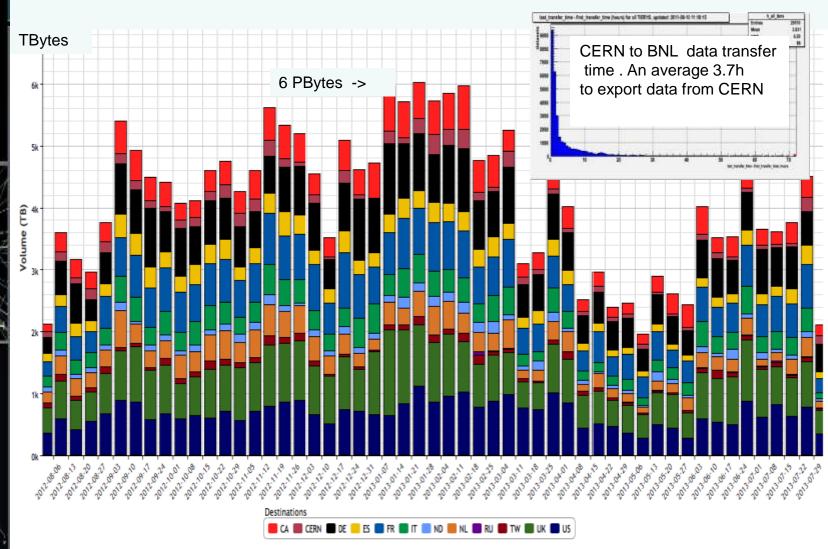
LHC Computing Grid: A global collaboration...





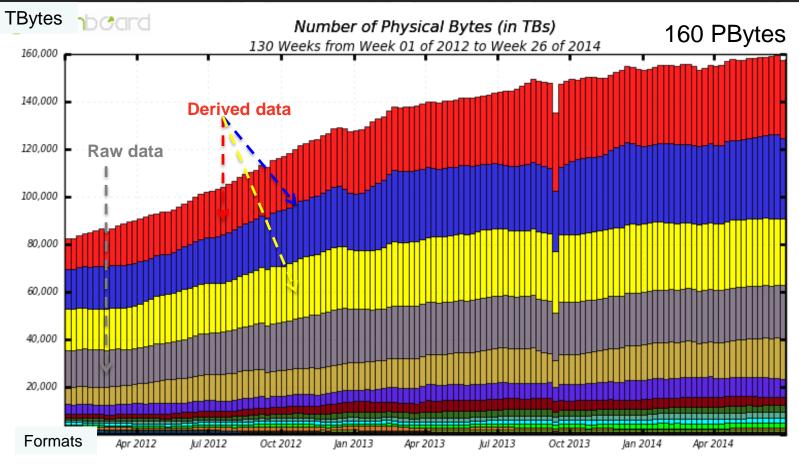
ATLAS Distributed Computing. Distributed Data Transfer

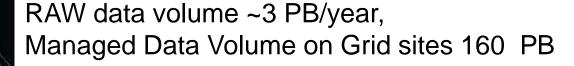
Data Transfer Volume in TB (weekly average). Aug 2012 - Aug 2013





ATLAS Data Volume on the Grid Sites







Processing the Experiment Big Data

- The simplest solution in processing LHC data is using data affinity for the jobs
 - Data is staged to the site where the compute resources are located and data access by analysis code from local, site-resident storage
 - However
 - In distributed computing environment we don't have enough disk space to host all our data at every Grid site
 - Thus we distribute (pre-place) our data across our sites
 - The popularity of data sets is difficult predict in advance
 - Thus computing capacity at a site might not match the demand for certain data sets
 - Different approaches are being implemented
 - Dynamic or/and on demand data replication
 - Dynamic : if certain data is popular over make additional copies on other Grid sites
 - On-demand: User can request local or additional data copy
 - Remote access: data can be accessed remotely
 - Both approaches have the underlying scenario that puts the WAN between the data and the executing analysis code
- Intelligent Workload Management System is needed to manage resources and to automate data processing, analysis and simulation



Workload Management System. Core Ideas

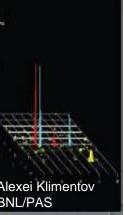
- Make hundreds of distributed sites appear as local
 - Provide central queue for users similar to local batch systems
- Reduce site related errors and latency
 - Build a pilot job system late transfer of user payloads
 - Crucial for distributed infrastructure maintained by local experts
- Hide middleware while supporting diversity and evolution
 - WMS interacts with middleware users see high level workflow
- Hide variations in infrastructure
 - WMS presents uniform 'job' slots to user
 - Easy to integrate grid sites, clouds, recently HPC sites
- Use the same system for Monte-Carlo simulation, data processing and users analysis

PanDA. Production and Data Analysis System





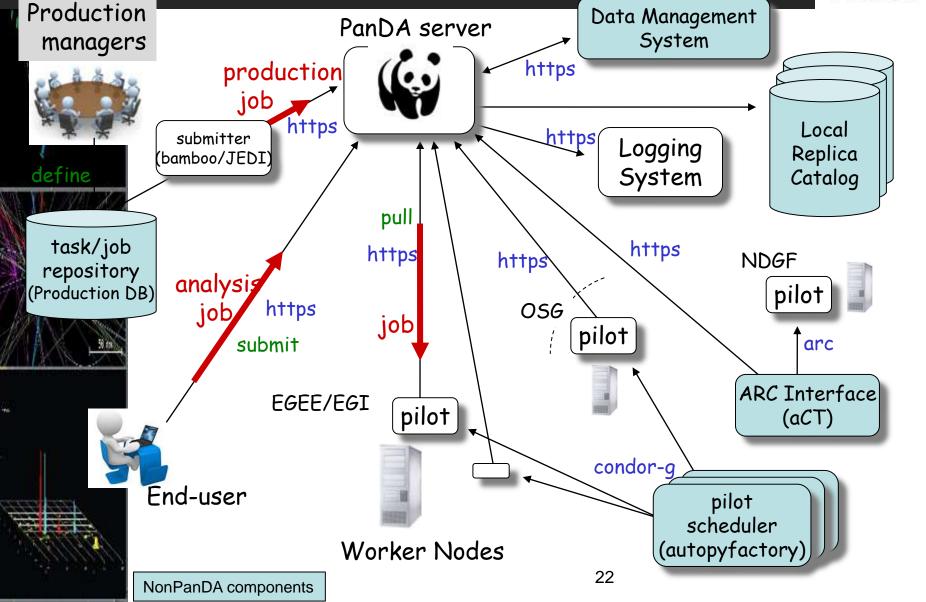




- ATLAS computational resources are managed by PanDA Workload Management System (WMS)
- PanDA project was started in fall of 2005 by BNL and University Texas at Arlington (UTA) groups
 - An automated yet flexible workload management system which can optimally make distributed resources accessible to all users
 - Standards based implementation
 - REST framework HTTP/S
 - Oracle or MySQL backends
 - About a dozen Python packages available from SVN and GitHub
 - Command-line and GUI/Web interfaces
- Through PanDA, physicists see a single computing facility that is used to run all data processing for the experiment, even though data centers are physically scattered all over the world.
- Now successfully manages O(10²) sites, O(10⁵) cores, O(10⁸) jobs per year, O(10³) users

Workload Management System

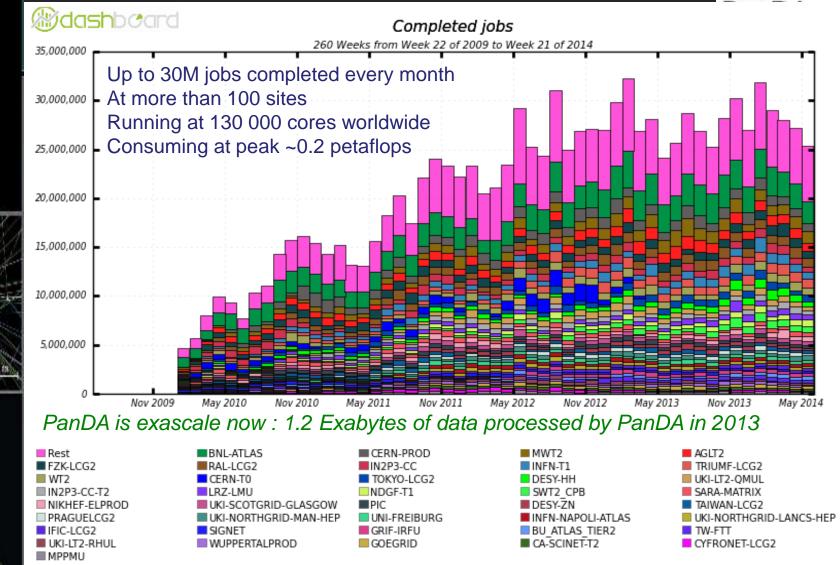




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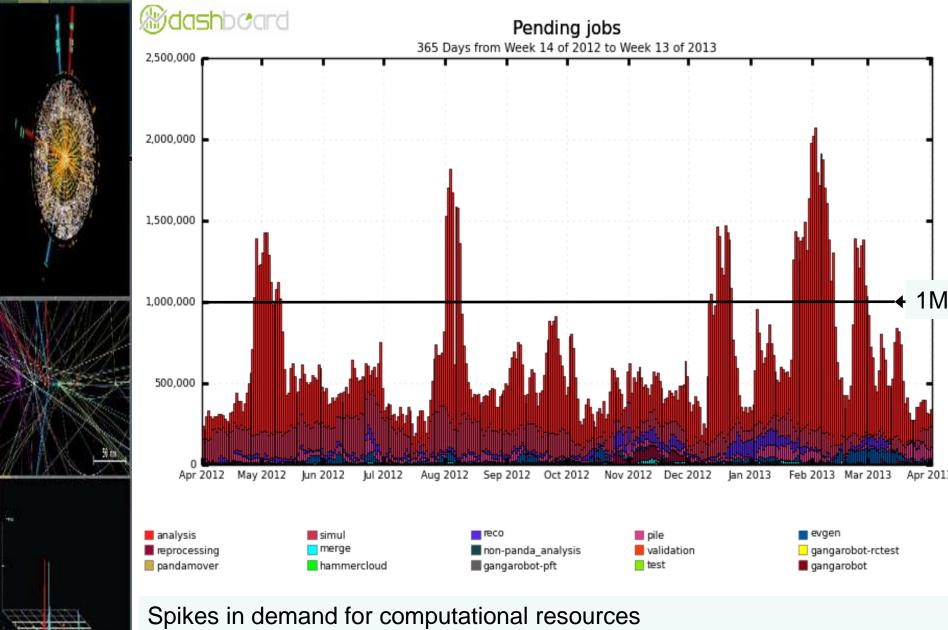
PanDA Performance





Maximum: 32,306,592 , Minimum: 0.00 , Average: 20,857,355 , Current: 25,377,087

All available resources are fully used



Spikes in demand for computational resources
Can significantly exceed available ATLAS Grid resources
Lack of resources slows down pace of discovery

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LHC Upgrade. Computing Needs

CPU needs per event

Run3

2019-2021

ATLAS

Run4

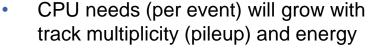
Run2:

2015 - 2017

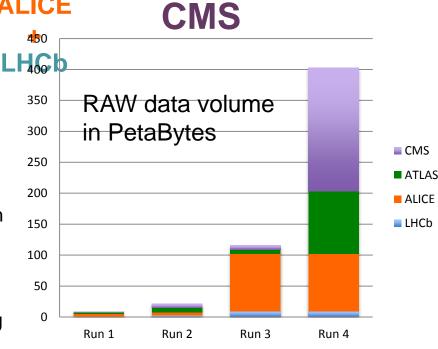
ALICE

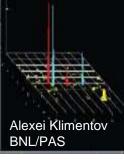
Run1:

2009 - 2013



- Storage needs are proportional to accumulated luminosity
- Grid resources are limited by funding and fully utilized







ATLAS Scale of Needs

- The ATLAS experiment uses a geographically distributed grid of approximately 130,000 cores continuously, to simulate, and analyse its data.
- The need for simulation and analysis would overwhelm the expected capacity of LHC Grid computing facilities unless the range and precision of physics studies were to be curtailed.
 - Physics requires to increase rate
 - Run1 data-taking rate 400 Hz
 - Run2 data-taking rate 1kHz
- Leadership Computing Facilities contributions of the order of 10 Million or more core hours per year become important and valuable.
- ATLAS computing can also be a close to ideal "crack-filling" application. PanDA WMS is being upgraded to make it aware of dynamically changing resources, and thus able to exploit groups of processors that become available for relatively short times.
- Extending PanDA beyond the Grid will further expand the potential user community and the resources available to them.



BigPanDA. Extending PanDA beyond the Grid



There are five dimensions to evolution of PanDA

- Making PanDA available beyond ATLAS and High Energy Physics
- Extending beyond Grid (Leadership Computing Facilities, Clouds, University clusters)
- Integration of network as a resource in workload management
- Integration of data management and workload management
- Using modern technologies such as non-relational databases

10/28/14



BigPanDA. Extending the scope. Cloud Computing.



Google Compute Engine (GCE) preview project: Google allocated additional resources for ATLAS for free: ~5M CPU hours, 4000 cores for about 2 months

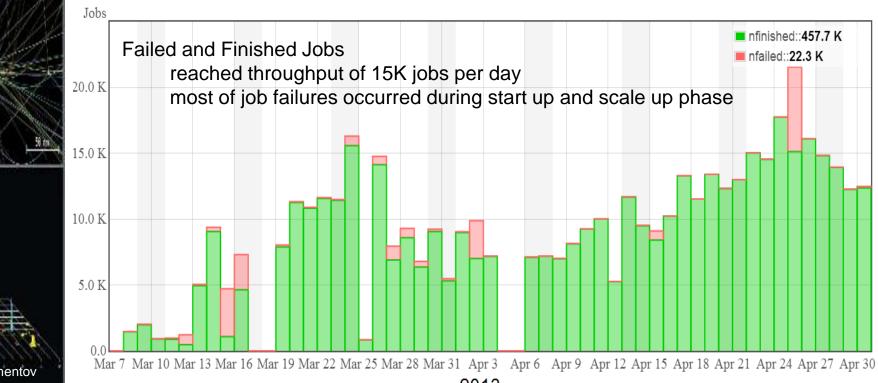
- ☐ Resources are organized as PanDA queues
- Output exported to BNL
- □ Transparent inclusion of cloud resources into ATLAS Distributed Computing
- □ The idea was to test long term stability while running a cloud cluster similar in size to Tier 2 center in US ATLAS
- □ Intended for CPU intensive Monte-Carlo simulation workloads, planned as a production type of run. Delivered to ATLAS as a resource and not as an R&D platform.

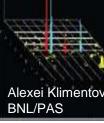


Running PanDA on Google Compute Engine



- We ran for about 8 weeks (2 weeks were planned for scaling up)
- Very stable running on the Cloud side. GCE was rock solid.
- Most problems that we had were on the ATLAS side.
- We ran computationally intensive jobs
 - Physics event generators, Fast detector simulation, Full detector simulation
- Completed 458,000 jobs, generated and processed about 214 M events
- Invited BNL PAS talk at Google IO conference







BigPanDA for Leadership Computing Facilities

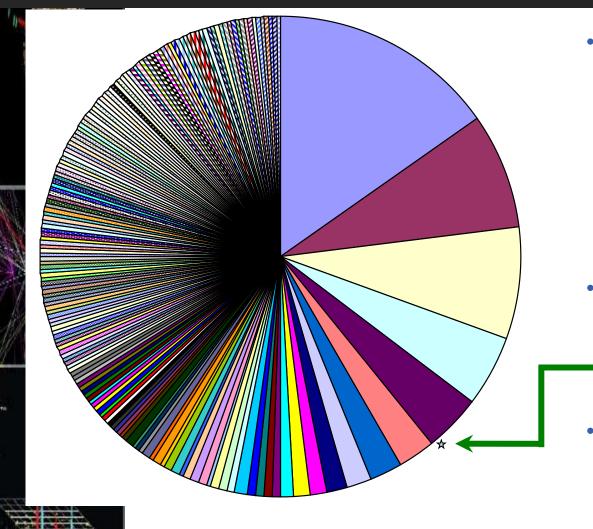


- Expanding PanDA from Grid to Leadership Class Facilities required significant changes in our system
 - Each LCF is unique
 - Unique architecture and hardware
 - Specialized Operating System, "weak" worker nodes, limited memory per worker node
 - Code cross-compilation is typically required
 - Unique job submission systems
 - Unique security environment
- HEP applications (such as Geant or ROOT) can effectively use a single core
- PanDA has potential to generate 300M hours per year
- PanDA submission algorithm has been adopted to use backfill information and to submit jobs in backfill mode.
- Final tests have been conducted in August together with OLCF team
 - Were able to collect ~ 200,000 core hours
 - Max number of nodes per job 5835 (93360 cores)
 - Close to 75% ATLAS Grid in size!

Used ~2.3% of all Titan core hours or ~14.4% of free core hours

The Top 500 Computers

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- Most of the computational power is concentrated in a small number of machines
 - Half the total power is in the top dozen computers
- Equivalent ATLAS
 Grid use is about the size of the sector
 (little green star)
- Equivalent ATLAS
 Grid would be around
 #27 on this chart

High Performance Computing (Top 10, Nov 2013)

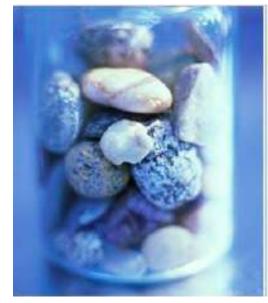
			cores	Rmax	Rpeak	Power
0	National Super Computer Center in Guangzhou China	Tianhe-2 (MilkyWay-2) - TH-IVB-FEP Cluster, Intel Xeon E5-2692 12C 2.200GHz, TH Express-2, Intel Xeon Phi 31S1P NUDT	3120000	33862.7	54902.4	17808
0	DOE/SC/Oak Ridge National Laboratory United States	Titan - Cray XK7 , Opteron 6274 16C 2.200GHz, Cray Gemini interconnect, NVIDIA K20x Cray Inc.	560640	17590.0	27112.5	8209
0	DOE/NNSA/LLNL United States	Sequoia - BlueGene/Q, Power BQC 16C 1.60 GHz, Custom IBM	1572864	17173.2	20132.7	7890
0	RIKEN Advanced Institute for Computational Science (AICS) Japan	K computer, SPARC64 VIIIfx 2.0GHz, Tofu interconnect Fujitsu	705024	10510.0	11280.4	12660
6	DOE/SC/Argonne National Laboratory United States	Mira - BlueGene/Q, Power BQC 16C 1.60GHz, Custom IBM	786432	8586.6	10066.3	3945
0	Swiss National Supercomputing Centre (CSCS) Switzerland	Piz Daint - Cray XC30, Xeon E5-2670 8C 2.600GHz, Aries interconnect , NVIDIA K20x Cray Inc.	115984	6271.0	7788.9	2325
0	Texas Advanced Computing Center/Univ. of Texas United States	Stampede - PowerEdge C8220, Xeon E5-2680 8C 2.700GHz, Infiniband FDR, Intel Xeon Phi SE10P Dell	462462	5168.1	8520.1	4510
B	Forschungszentrum Juelich (FZJ) Germany	JUQUEEN - BlueGene/Q, Power BQC 16C 1.600GHz, Custom Interconnect IBM	458752	5008.9	5872.0	2301
0	DOE/NNSA/LLNL United States	Vulcan - BlueGene/Q, Power BQC 16C 1.600GHz, Custom Interconnect IBM	393216	4293.3	5033.2	1972
10	Leibniz Rechenzentrum Germany	SuperMUC - iDataPlex DX360M4, Xeon E5-2680 8C 2.70GHz, Infiniband FDR IBM	147456	2897.0	3185.1	3423

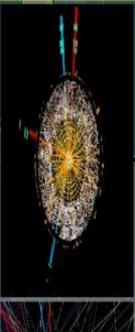
[★] Collaborations have members with access to these machines and to ARCHER, NERSC, ... Some HPCs are already successfully used as part of Nordu Grid (Abisko, Abel, Triolith, C2PAP, Hydra)



HPC Scheduling

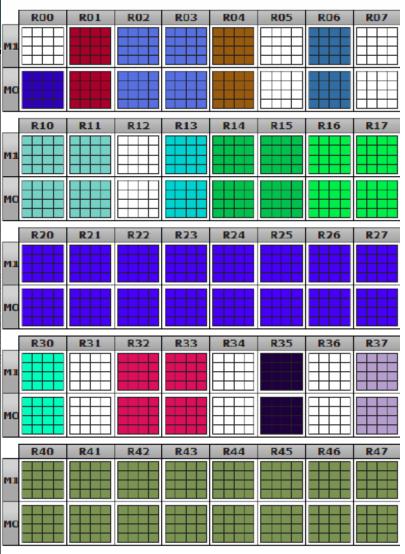
- HEP applications (such as Geant or ROOT) can effectively use a single core
- HPC is full, means that the system have allocated all the cycles it is able to deliver
 - It is probably not all cycles it has
 - Just as there is room for sand in the jar of rocks, there's room for HEP jobs on even a "full" HPC





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HPC Scheduling. Cont'd



- This is not a typical workday view of HPC machine
- At this moment machine is 85% full, the largest open partition has 1024 nodes
- But the shortest job in the queue required 4096 nodes
- The scheduler will be happy to run a short job in R12
- 24h backfill tests have been conducted on Titan together with OLCF team



Free nodes



27 PFlops (Peak)

18,688 compute nodes with GPUs

299,008 CPU cores

AMD Opteron 6200 @2.2 GHz (16 cores)

32 GB Ram per node

Nvidia TESLA K20x GPU per node

32 PB disk storage (Luster)

29 PB HPSS tape archive



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Interfacing PanDA with Titan



- BigPanDA project on Titan under ASCR auspices
- 10M hours allocation for 2014-15 on Titan
 - Access to EOS new Cray XC30 machine at OLCF
 - Also we have access to NERSC via OSG and ATLAS allocations
- Collaboration between ATLAS, ALICE, nEDM experiments
- Project members from CERN, BNL, UTA, ORNL, UTK, LBNL, ...
 - Strong interest from OLCF, took responsibility for MPI wrapper base and docs
- PanDA has potential to generate 300M hours per year
- Technology developed on Titan should be applicable for other HPC centers
 - Functionality tests have been conducted on NERSC
 - Interest from ASGC, NRC-KI, JINR, Ostrava

2/9/2014 ACAT 2014



Interfacing PanDA with Titan. Cont'd

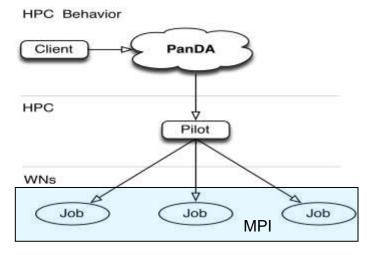


- PanDA modular pilot augmented with HPC specific classes
- SAGA (Simple API for Grid Applications) framework as pilot's interface to HPC batch schedulers
 - http://saga-project.github.io/saga-python/
 - http://www.ogf.org/documents/GFD.90.pdf
- MPI wrapper/overlay scripts that allow to run multiple single threaded workload instances in parallel
- "Backfill" functionality in pilot

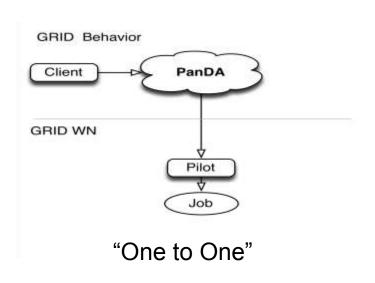
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Pilot on HPC with MPI wrapper



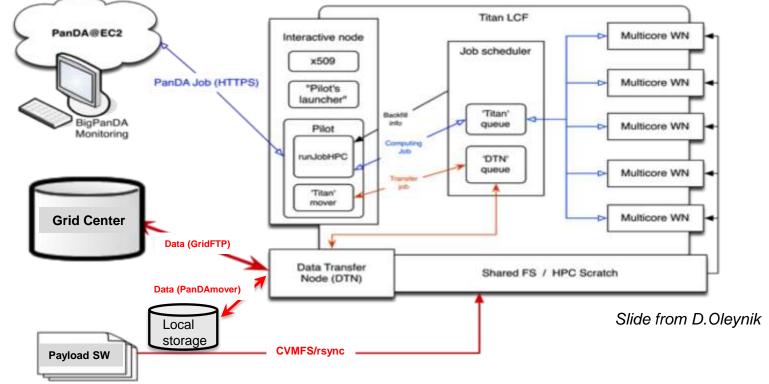
"One to Many"





Extending PanDA to Oak Ridge Leadership Computing Facilities



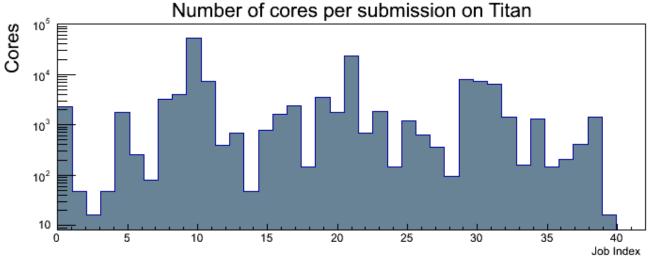


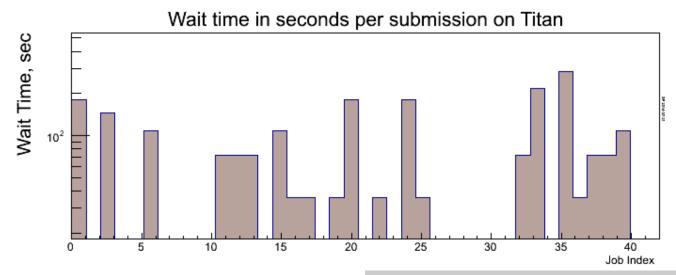
- ATLAS (BNL, UTA), OLCF, ALICE (CERN, LBNL, UTK):
 - adapt PanDA for OLCF (Titan)
 - reuse existing PanDA components and workflow as much as possible.
 - PanDA connection layer runs on front-end nodes in user space. There is a predefined host to communicate with CERN from OLCF, connections are initiated from the front-end nodes
 - SAGA (a Simple API for Grid Applications) framework as a local batch interface.
 - Pilot (payload submission) is running on HPC interactive node and communicating with local batch scheduler to manage jobs on Titan.
 - Outputs are transferred to BNL T1 or to local storage
- The same architecture has been tested on other super-computers



PanDA pilot tests on Titan







Average wait time 70 seconds!



Resources Accessible via PanDA





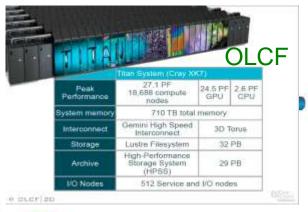


























Summary

- 2000s The decade of the Grid
- First years of LHC data Distributed Computing has helped deliver physics rapidly
- Entering a phase of computing evolution
- Challenges for computing scale & complexity will continue to increase dramatically
- The distributed computing model allows us to incorporate clouds and LCF/HPC centers and to use them efficiently for LHC Run 2 and beyond
- Access to the LCF coupled with collaborative help in the transformation of HEP code would be a major scientific contribution to the physics discoveries of the next ten years
- The work on extending PanDA to Leadership Computing Facilities has started. PanDA has been successfully ported on OLCF Titan and NERSC, and it is underway to the supercomputer at NRC-KI



Summary. PanDA's Success



- PanDA was able to cope with increasing LHC luminosity and ATLAS data taking rate
- Adopted to evolution in ATLAS computing model
- Several leading HENP and astro-particle experiments (ALICE, AMS, LSST, COMPASS) has chosen PanDA as workload management system for data processing and analysis or evaluating it.
- It is also used for Biomedical applications
- Cost effectiveness. PanDA enables BigData processing at a fraction of the cost at Google/Amazon data centers

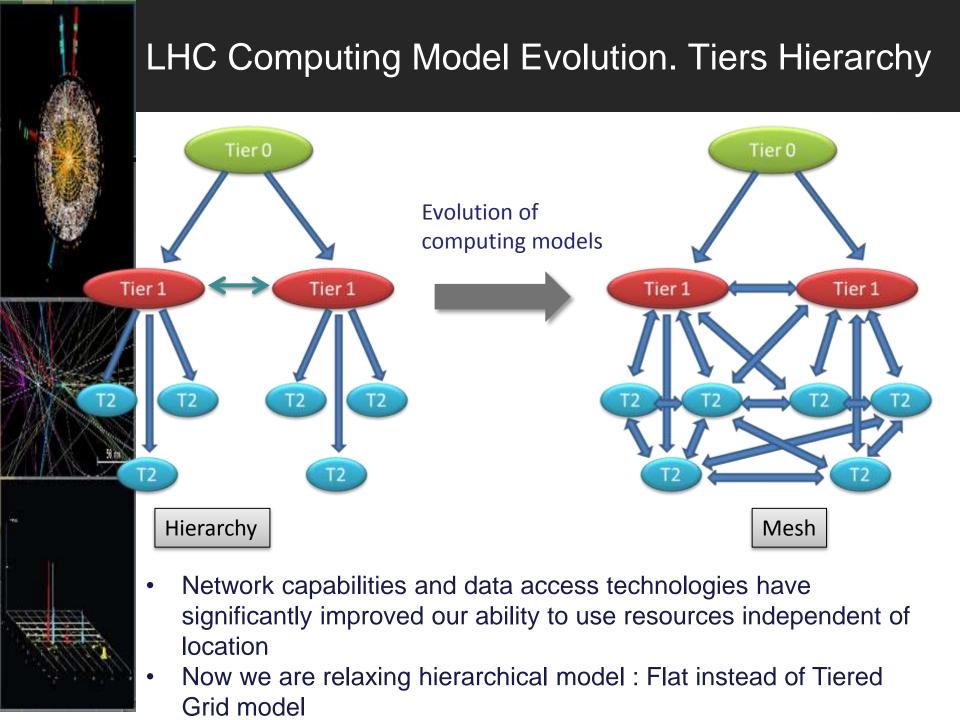
PanDA is exascale now: 1.2 Exabytes of data processed by PanDA in 2013



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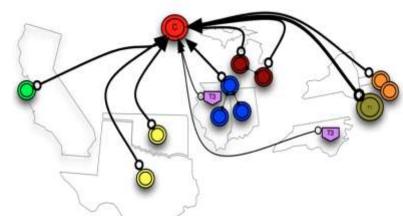
Backup slides







LHC Computing Model Evolution. Global Data Federations... (Reducing Complexity)



ATLAS: Global Data Federation - FAX

ALICE:

- Virtually joining together the sites based on proximity (latency) and network capacity into Regional Data Clouds
- Each cloud/region provides reliable data management and sufficient processing capability
- Dealing with handful of clouds/regions instead of the individual sites

CERN-IT: site or cloud is represented by Virtual Storage Cloud Node

