## An Infrastructure for (High Performance) Data Analysis at DESY

- > Current status
- > Plans for further development
- > Conclusions

Sergey Yakubov, Stefan Dietrich, Uwe Ennslin, Jürgen Hannappel, Janusz Malka, Cartsen Patzke, Birgit Lewendel, Martin Gasthuber DESY IT Hamburg, 25 January 2018





- > High rate detectors
- > May run in parallel

#### > Next generation will come

Detector	OS/Access	File size/rate	Bandwidth
Pilatus 300k	Linux (Black box)	1,2 MB Files @ 200 Hz	240 MB/s
Pilatus 6M	Linux (Black box)	25 MB files @ 25 Hz 7 MB files @ 100 Hz	625 MB/s 700 MB/s
PCO Edge	Windows	8 MB files @ 100Hz	800 MB/s
PerkinElmer	Windows	16 MB + 700 Byte files @ 15 Hz	240 MB/s
Lambda	Linux	60 Gb/s @ 2000 Hz	7.5 GB/s
Eiger	Http (Black Box)	30 Gb/s @ 2000 Hz	3.8 GB/s

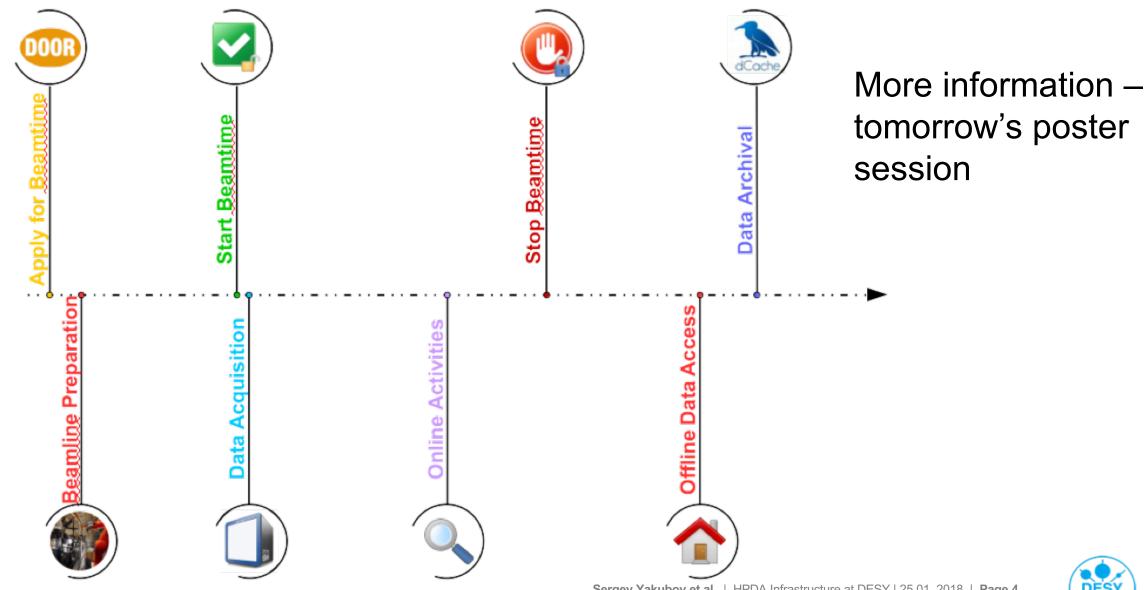




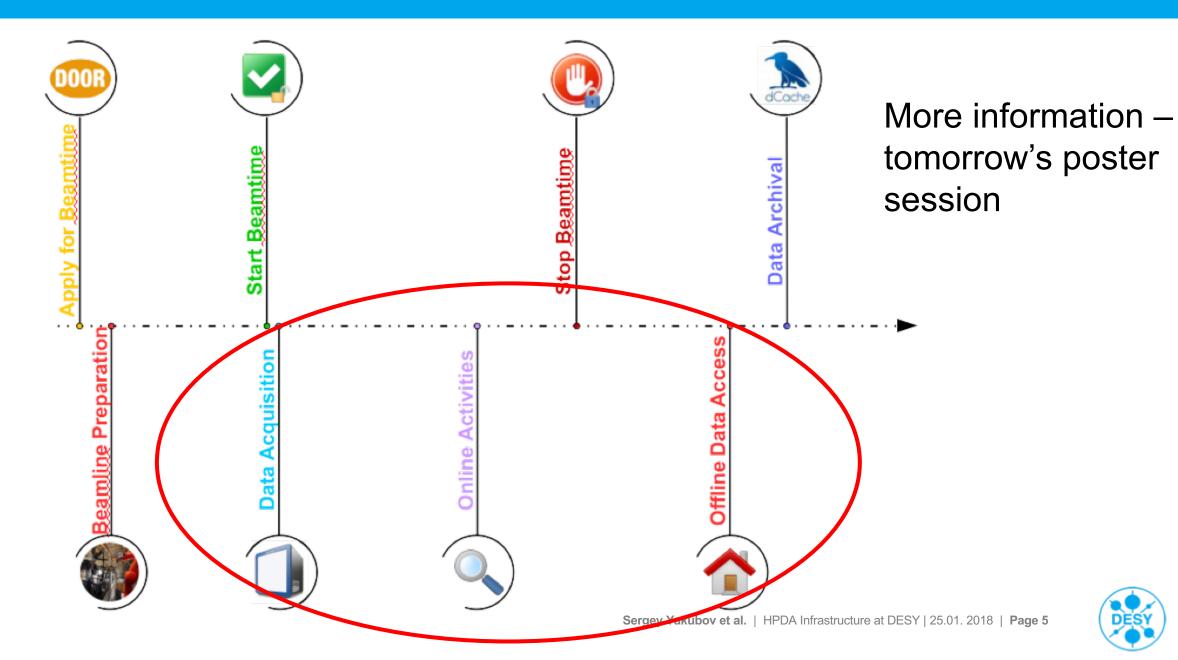
# Key assumption – data is stored and analysed in a central compute center





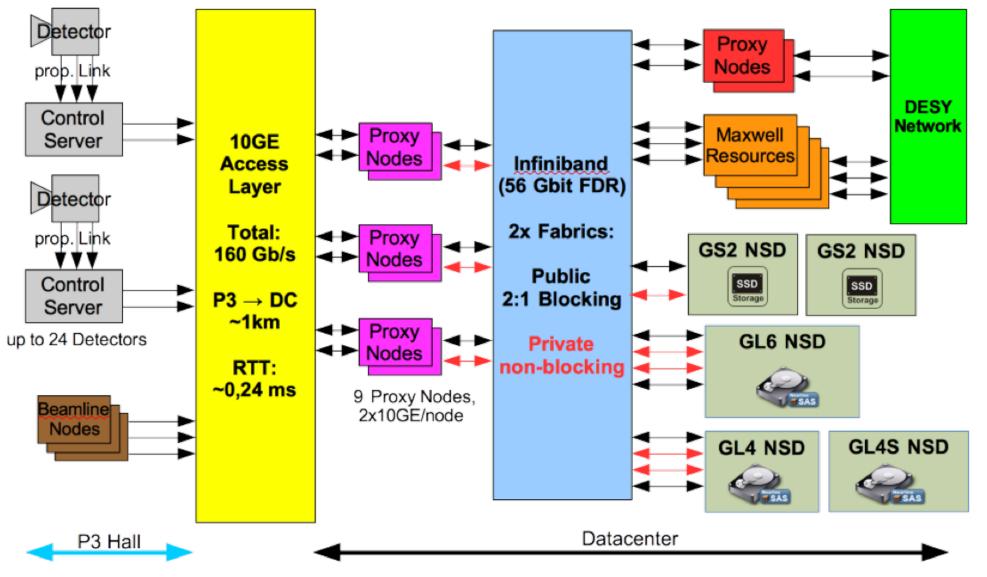


#### **ASAP<sup>3</sup>**

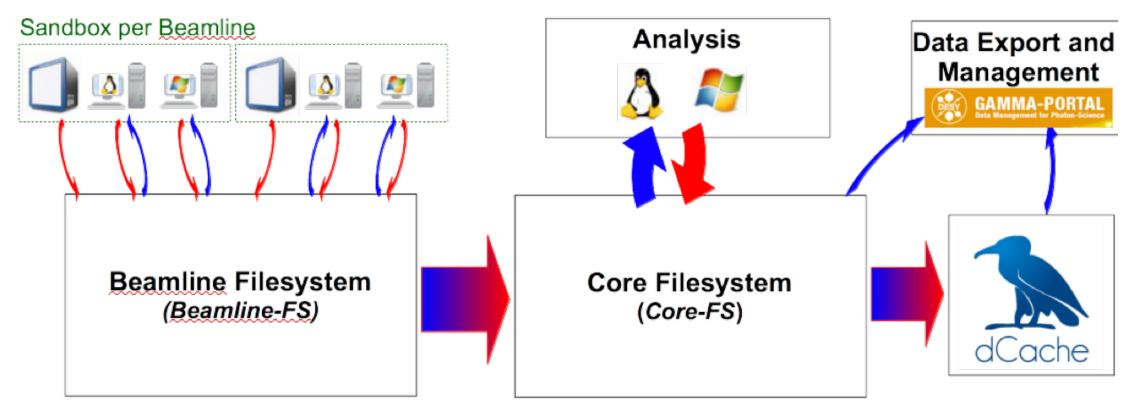




#### **ASAP<sup>3</sup> – Hardware Architecture**





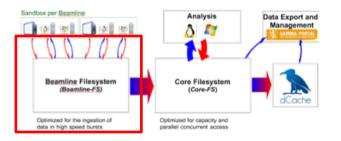


Optimized for the ingestion of data in high speed bursts

Optimized for capacity and parallel concurrent access

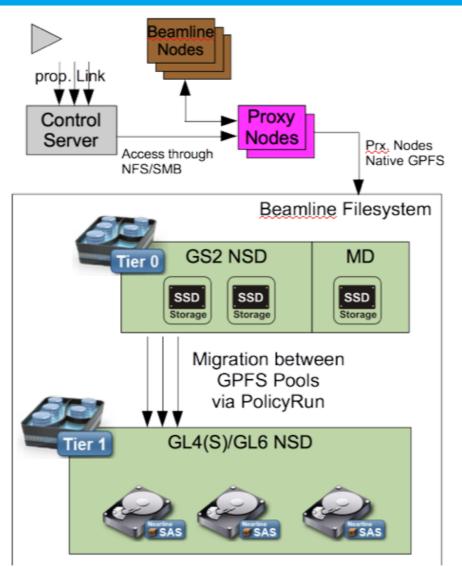


## ASAP<sup>3</sup> – Beamline filesystem



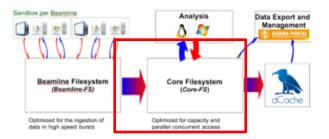
- > Only host based authentication, no ACLs
- > Access through NFSv3, SMB
- > Optimized for performance
  - NFSv3: ~600 MB/s
  - SMB: ~300-600 MB/s
- > Tiered Storage
  - Tier 0: SSD burst buffer (<10 TB)</p>
  - Migration after short period of time
  - Tier 1: ~90 TB capacity

Limited/no access for user analysis jobs via native GPFS



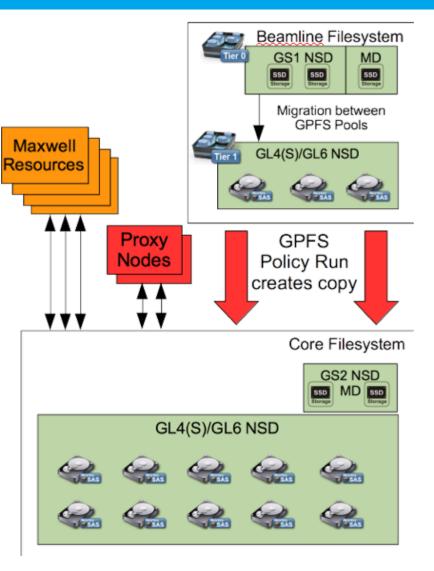


## ASAP<sup>3</sup> – Core filesystem



- Full user authentication
- > Access through NFSv3, SMB or native GPFS
- > GPFS Policy Runs copies data
  - Beamline → Core Filesystem
  - Single UID/GID
  - ACL inheritance gets active
- > 2 snapshots per day

Full access for user analysis jobs (delay 4 min)



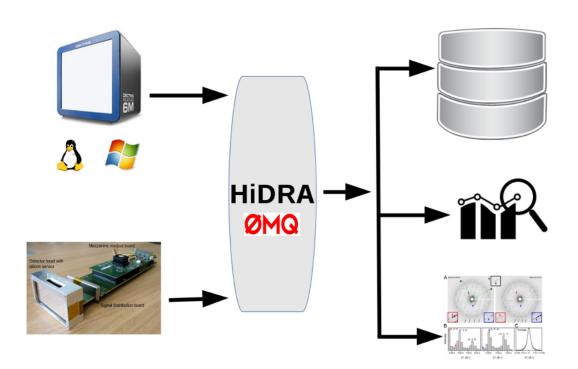


		Bandwidth
Detectors	Pilatus 300k	240 MB/s
	Pilatus 6M	700 MB/s
	PCO Edge	800 MB/s
	PerkinElmer	240 MB/s
	Lambda	7.5 GB/s
	Eiger	3.8 GB/s
Experimental Hall to Maxwell (aggregate)		20 GB/s
NFS,SMB		600 MB/s



## **HiDRA\***

- Generic tool set for high performance data delivery
- Based on Python and ZeroMQ
- Messaging system
  - Push to subscribers
  - Request-reply
- > Various use cases
  - Storing data in filesystem
  - Online analysis/monitoring



## \* Talk from Manuela Kuhn



## Status as of 2017- present

		Bandwidth	
	Pilatus 300k	240 MB/s	
Detectors	Pilatus 6M	700 MB/s	
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Experimental Hall to Maxwell (aggregate)		20 GB/s	
NFS,SMB		600 MB/s	
HiDRA		<b>1.2 GB/s – now</b> 1.2 GB/s x N – in development	

DES

#### Scalability

- Producer network scaling: Nx10GE, Nx40GE (multithreading Python?)
- Workers able to start large number (100-1000) of workers

#### > Higher transfer rates

- Multiple links
- RDMA over Infiniband or RDMA over Ethernet (ZeroMQ?)
- Decouple data taking and data processing
  - Variable injection/data processing rate
  - Support for offline analysis

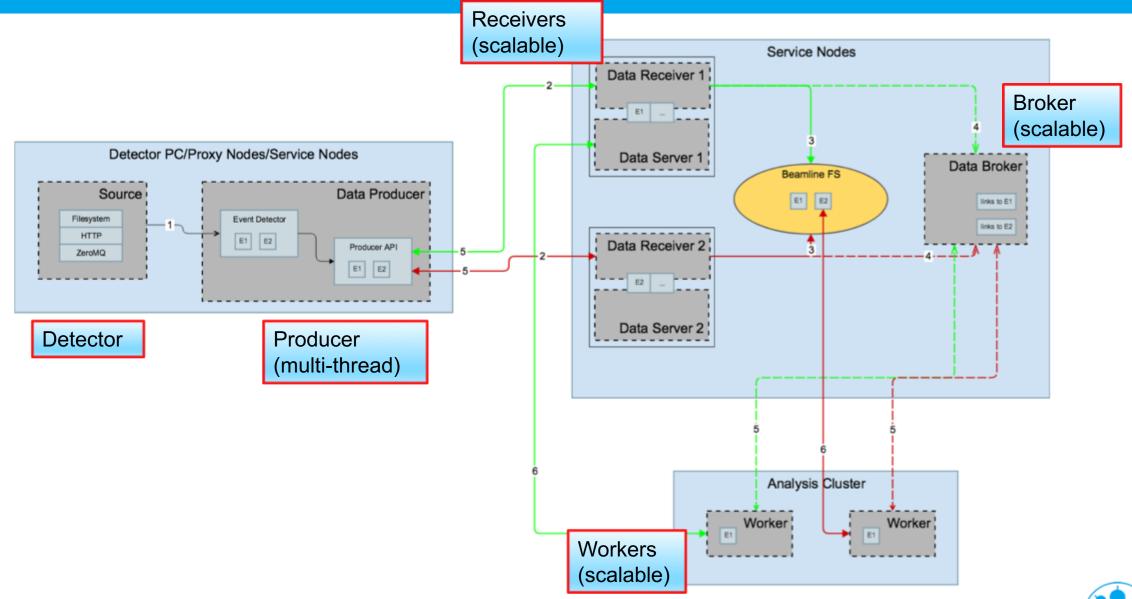


## **ASAP::O - Addressing the Challenges**

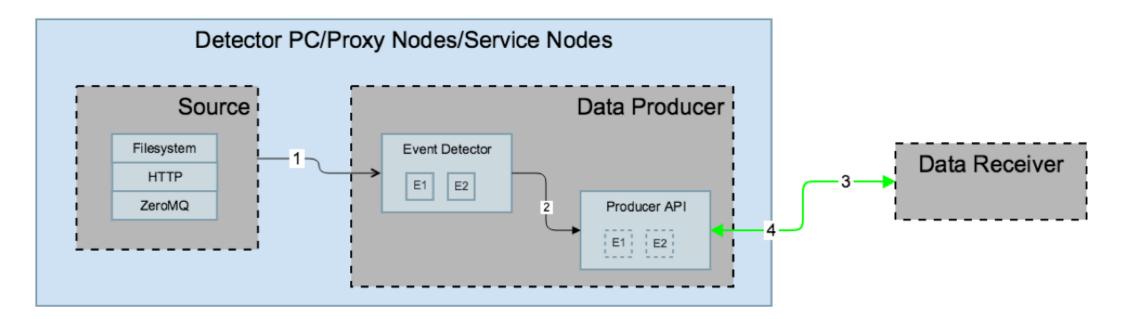
- > A new middleware platform for high-performance next-generation detector data analysis
- Supports data analysis synchronous and asynchronous to data taking
- Scalable (N detectors, K network links, L service nodes, M analysis nodes)
- > Highly available (services in Docker containers managed by Kubernetes)
- Efficient (C++, multi-threading, RDMA, ...)
- Provides user friendly API interfaces (C/C++, python, REST API)
- Supports various platforms and OS
- Compatible with HiDRA API



#### **ASAP::O - Architecture**







- Event Detector to be written for various detectors
- > Python interface to be discussed :)
- Interface to HiDRA

Get data from a detector and send it to the Receiver using Producer API

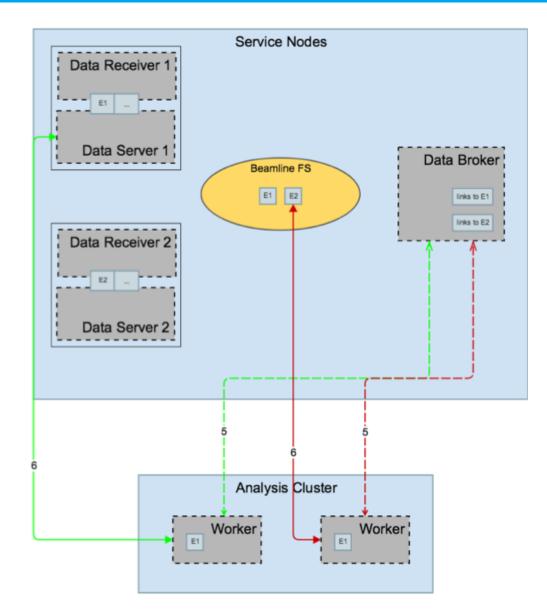


#### Location

- Detector (blackbox) writes locally to the NFS/SMB mount point
- Detector writes/sends data to the detector's PC
- Detector (blackbox) sends data to proxy node
- Data granularity
  - Sub-images
  - Single images
  - HDF5 files with multiple images
- > Detector/user metadata
  - No metadata
  - Metadata is periodically written



## ASAP::O - Worker



- Request/Reply pattern
- Same user code for offline/online
- > Python interface will be provided
- Can/will run in isolated Docker containers on Maxwell nodes of certain type (memory, GPU,...) with access to Beamline FS
- Interface to HiDRA

Use Worker API to retrieve data



## **ASAP::O - Worker Use Cases**

#### Software

- User code (Linux, Windows), C++/Python
- MatLab, IDL, commercial applications
- reduced performance possible

#### > Analysis

- online
- offline
- testing (online with virtual detectors/offline)
- at 3rd party sites/at home
- Parallelization
  - All images can be processed independently
  - Sequential processing (with chunks of N images)

#### > HDF5 writer

- pack images into HDF5 files
- add metadata to HDF5 files
- > Using filters/queries
  - process only images with a specific condition (from time A to time B, from frame N to frame M, with metadata X, ...)
- Multi-stage processing/pipelining
  - Worker as Producer
  - Connectors to Apache Storm, …



## **ASAP::O - Current Status and Development Process**

- > Work in progress (first version this year)
- > Agile project (almost)
  - Following Kanban approach, without time tracking
- Using Atlassian Tools provided by DESY IT for CI/CD.
  - Confluence for documentation (development workflow, code conventions, design sketches, ...)
  - Bitbucket for source code
  - Jira for task tracking
  - Bamboo for testing/deployment
- Heavily tested
  - Unit tests (sometimes written first)
  - Integration tests
  - Test coverage close to 100%
  - Memory leaks testing
  - Testing on both Windows and Linux



#### Conclusions

- Data analysis infrastructure in current state supports high rate data transfer to the compute center and is able to support online data analysis
- > A new platform ASAP::O is being developed
  - Even higher rate data transfer to beamline and core filesystem
  - High performance online and offline data analysis using same user code
  - Transparent access to images (no need to know directory structure, file names)
  - Various modes to access images: last image, arbitrary image, ordered sequence, in parallel
  - Select images based on metadata
  - Multi-stage processing of data streams and connections to other datastream processing frameworks
  - Prepare everything at home, deploy on Maxwell (e.g. Docker)

· ...

Close(r) coordination with users is essential for success. Participation in development is possible/welcome.

