



International School of Trigger and Data Acquisition





## Design and Implementation of a Monitoring System

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## What you are expected to learn in the next hour

- Why systems need to be monitored
- □ A little bit of theory
- The Basic one-size-fits-all Architecture
  - Technology independent
- □ Implementation Strategy:
  - With a few technology examples
- Data Quality Monitoring



#### Why systems need to be Monitored?

- Cos the Universe is not Perfect
- The rate of failures is proportional to the system complexity
- Monitoring is indispensable for the System control



## How Higgs boson discovery would have looked like in an ideal world

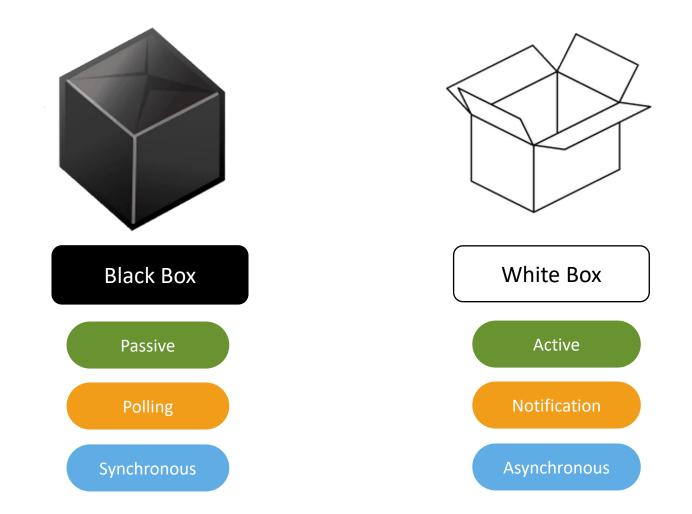


#### What happens in reality

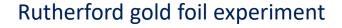


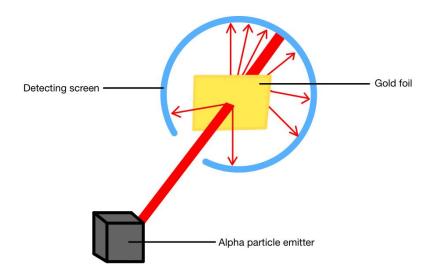
- A complex project has a chance to success only if it is ready to deal with problems
- Monitoring System provides the first line of defense:
  - Detects, Reports, Helps to Investigate

#### Two Main Approaches for Monitoring



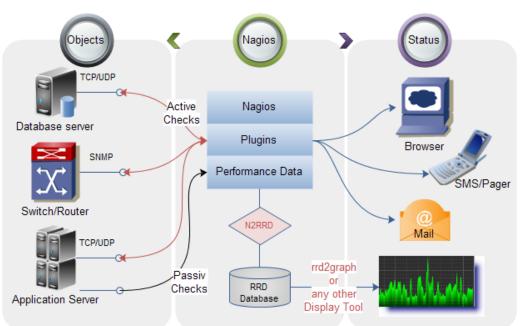
#### The Black Box Monitoring Approach





- System to be monitored is a *Black Box*
- Use well-defined procedures as probes for the *Black Box* and measure the result

#### The **Black Box** Monitoring Example



- <u>Nagios</u> is a classical example of the *Black Box* monitoring
  - Provides checks for commodity HW and SW
  - Allows to integrate custom checks
- Other examples: <u>lcinga</u>, <u>Ganglia</u>, etc.

#### Black Box Approach for DAQ system?

- Data AcQuisition is an heterogeneous field
  - Boundaries not well defined
  - An alchemy of physics, electronics, networking, computer science, ...
  - Hacking and experience
  - ..., money and manpower matter as well

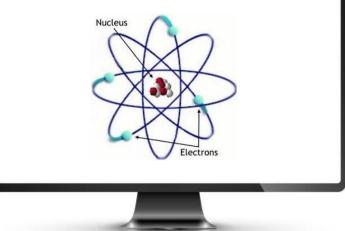


- DAQ system components operate at high rate:
  - Polling for monitoring information is inefficient
- A DAQ system has many **custom** HW and SW:
  - Good opportunity to do monitoring in a better way...

## What if the Universe was created by a Computer Scientist?



#### **The White Box Approach**

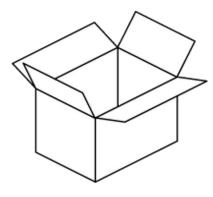


- Objects expose information about their states:
  - E.g. coordinates and velocities of the particles
- Physicists would merely take care of visualizing this information

## Off-the-shelf Software Solutions?



Plenty of ready-made solutions Only applies to commodity HW/SW Custom components require development of custom probes



No ready-made solutions

- Can be constructed using:
- Commodity tools and libraries
- A little bit of custom programming
- Some good recipes

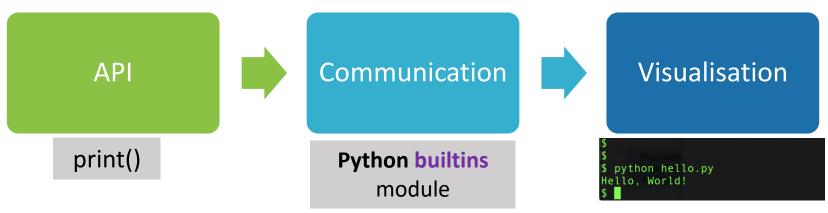


### The Simplest "White Box" Example





#### The Basic Architecture



#### • API

- The critical component
- DAQ system see only this API
- It must be independent of the **Communication** and **Visualisation**
- This is not the case!
  - Communication and Visualisation are strictly bound to the print() function API

#### Another Issue with the print() function

#### print("Hello, World")

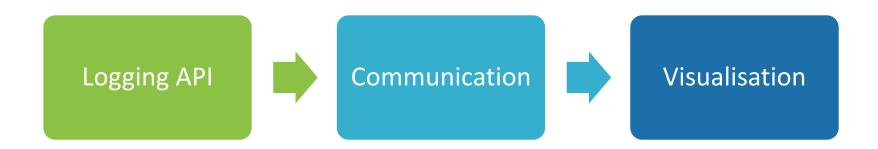
- Not bad for a single application but doesn't scale
- With multiple applications running for a long time, we want to know:
  - When did something happen?
  - Where did it come from?
  - How important is it?
- Do better solutions exist?



### Logging API to the rescue



#### The Basic Architecture



- The Logging API and Communication layers are fully independent
- The Logging API
  - Well-designed and mature
- Communication:
  - Different implementations exist on the market
  - Can be exchanged transparently for the end-user applications

## Programming Languages Support

#### Python

import logging

#### class Logger:

def critical(msg, \*args, \*\*kwargs):

def debug(msg, \*args, \*\*kwargs):

def error(msg, \*args, \*\*kwargs):

def info(msg, \*args, \*\*kwargs):

def warning(msg, \*args, \*\*kwargs):

#### Java

import java.util.logging.Logger

#### class Logger {

void severe(String msg);

void fine(String msg);

void error(String msg);

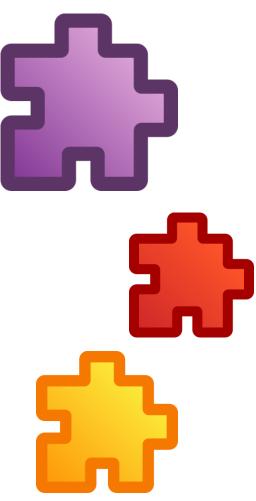
void info(String msg);

void warning(String msg);

}

### Existing Appenders for the Java Logging API

- **CassandraAppender** writes its output to an <u>Apache</u> <u>Cassandra</u> database
- FileAppender writes events to an arbitrary file.
- FlumeAppender <u>Apache Flume</u> is a distributed, reliable and highly available system for efficiently collecting, aggregating, and moving large amounts of log data
- JDBCAppender writes log events to a relational database table using standard JDBC
- NoSQLAppender writes log events to a NoSQL database
- **SMTPAppender** sends an e-mail when a specific logging event occurs, typically on errors or fatal errors
- ZeroMQAppender uses the <u>JeroMQ</u> library to send log events to one or more ZeroMQ endpoints



#### What about C++?

Rare case where using MACRO for the public API is a viable option

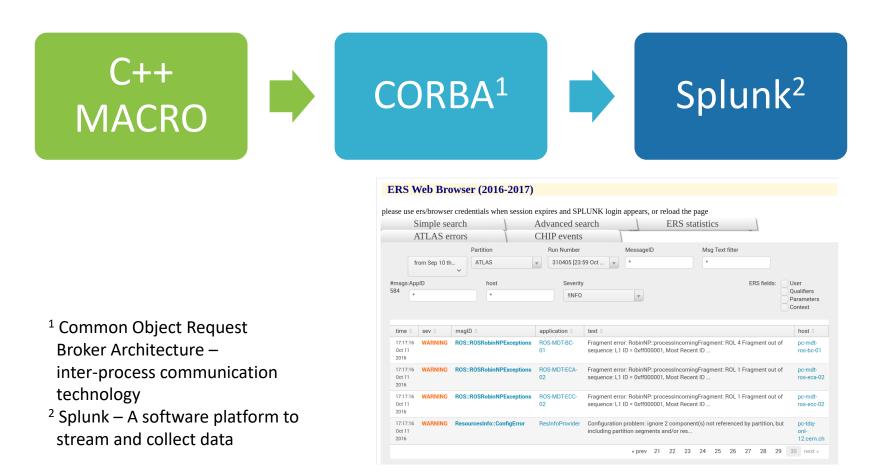
DAQ\_LOG\_CRITICAL("File '" << file\_name << "' not found")
DAQ\_LOG\_ERROR(...)
DAQ\_LOG\_WARNING(...)
DAQ\_LOG\_INFO(...)
DAQ\_LOG\_DEBUG(...)</pre>

• Initial implementation may be trivial:

#define DAQ\_LOG\_CRITICAL(m) std::cerr << m << std::endl;</pre>

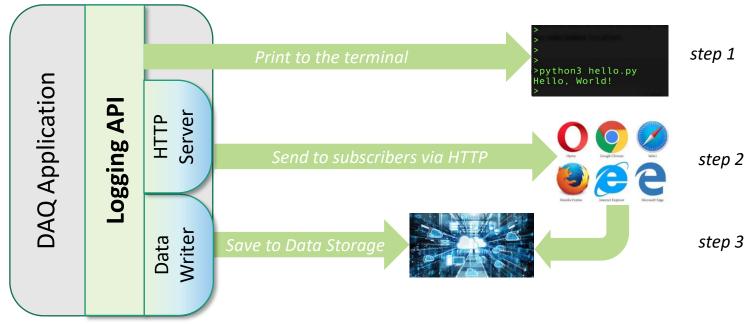
- A scalable implementation can be provided later:
  - Will not affect users' code

## The ATLAS Experiment: Error Reporting System



#### Evolving the Monitoring System Implementation

- The destination of the messages can be changed at any moment:
  - No changes in the Software Applications required!
- Data Storage is optional but very handy:
  - Adds **persistence** can be used for postmortem analysis

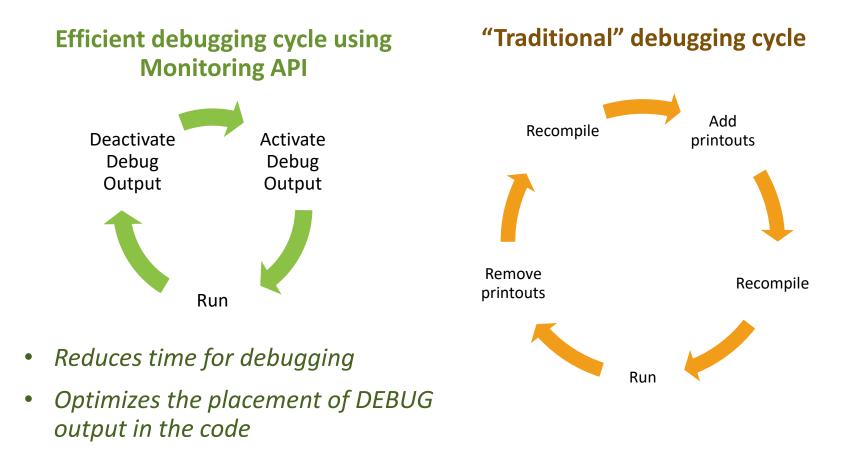


#### Set Priorities Properly

- Choose (or implement) the Monitoring API before starting to implement the DAQ system:
  - The Monitoring must be used by all components of the DAQ system
  - Changing them later will be a pain
- Can take care about Communication and Visualization implementations later:
  - Using simple output to terminal would be sufficient for the beginning
- Advantages:
  - Using the monitoring system will exercise its functionality and performance
  - Learn the best ways of presenting information
  - Speed up the DAQ system development



#### How Monitoring System can speed up DAQ System Development



Can we Extend the Same Ideas to the Other Types of Monitoring Data?

## Monitoring Data Types

- Messages used to inform about anything of importance that happens in the system
- **Metrics** show how the system performs:
  - Values of properties of the software and hardware system components



## Main Metrics Types

#### Counter



- Monotonically increasing integer number
- Simple to monitor:
  - Last value for the last time period
- Examples:
  - Cumulative totals: number of triggers, number of bytes sent/received, etc.

#### Gauge

- Arbitrary changing value:
  - Integer or floating point
- Monitoring can be tricky:
  - Last value
  - Mean value
  - Min/Max values
- Examples:
  - Resources usage: CPU, memory, buffe
  - Rates: triggers/s, bytes/s, etc.
  - HW Properties: voltage, current, temperature, etc.



#### Metrics Monitoring Requirements





# ✓ Must be displayed as time series ✓ Must be accessible in real-time ✓ Must be recorded to be checked later

#### Reusing the Same Architecture

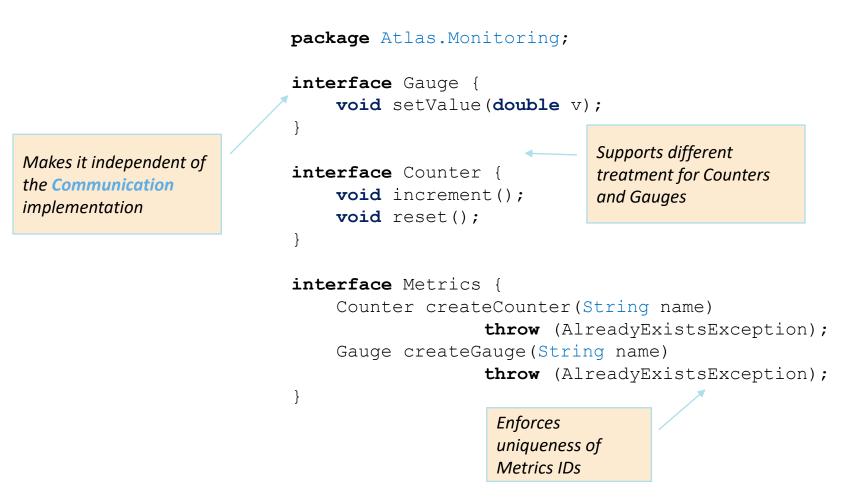


- API must be independent of the Communication and Visualisation
- Communication and Visualisation may be changed many times during the project life-time

#### A Common API for Metrics?

- There is no commonly accepted API for Metrics:
  - SW tools for metrics collection and analysis use their proprietary APIs
- This may not be a problem for a small short-living project:
  - Directly using a specific SW API is a viable option
  - Be careful to choose a SW with the live-time going beyond your project time-scale
- HEP experiments have a life-time of O(10) years:
  - It's difficult to find a SW system that is likely to survive that long

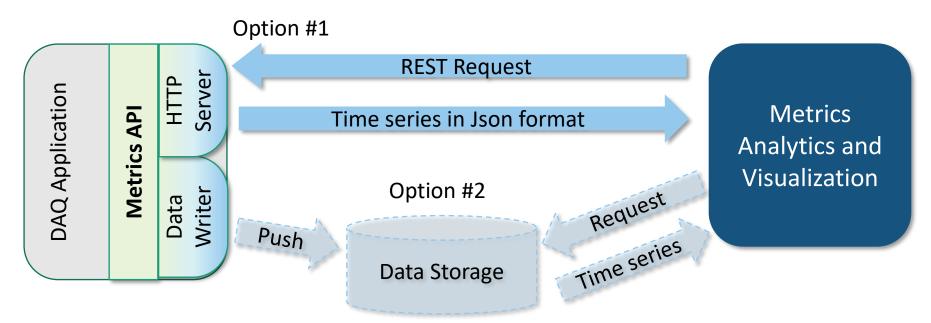
#### Custom API for Metrics Monitoring



#### Metrics IDs

- All Metrics must have unique IDs
- Uniform naming schema greatly simplifies Metrics handling:
  - Finding required Metrics is straightforward
  - Easy selection and filtering using regular expressions
- A possible approach:
  - System/Sub-system/Component/ + Metrics Name
- Examples:
  - /DAQ/DataFlow/EventRecoder/EventsNumber
  - /DAQ/DataFlow/EventRecoder/RecordingRate

## Some Implementation Options



- The underlying implementation can be updated as the main project evolves:
  - Does not affect the DAQ applications
  - The same Analytics and Visualization tools can still be used

#### **RESTful Protocol**

- REST **Re**presentational **S**tate **T**ransfer
- Client-server HTTP-based stateless communication protocol
- Supported by most of the modern information storage as well as Web-based Visualisation systems:
  - Supports seamless interoperations
- Makes it easy to switch from one Storage or Visualisation platform to another

#### **REST Protocol Example**

#### • Request:

https://atlasop.cern.ch/monitoring/

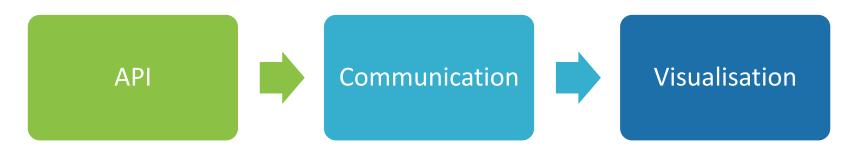
- ? id=ATLAS.Dataflow.RecordedEvents.Rate
- & from=now-30d
- & to=now
- Response:

```
Json Time Series, e.g.:
[
    {t:1579104640,v:12345},
    {t:1579104645,v:12346},
    {t:1579104650,v:12347},
    {t:1579104655,v:12348}
```

#### Web-Based Visualization Tools

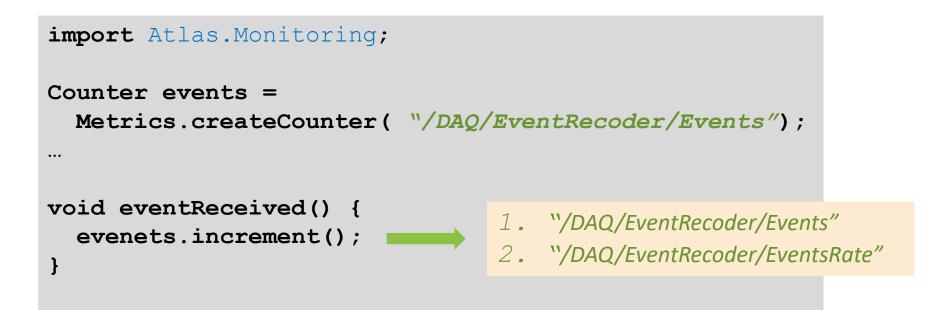
- Javascript tools which work in Web Browsers:
  - Grafana the open observability platform
  - D3 a low-level JavaScript toolbox for data visualization
  - Rickshaw a JavaScript toolkit for creating interactive time series graphs
  - There are many others as well...
- Very convenient for the end users:
  - Don't require extra software installation
  - Provide real-time monitoring data access from any place of the World

#### Are there some other Advantages?



- The API can hide implementation of common data handling patterns
- Produce Derivative Metrics
- Perform Metrics Rate Down-sampling
- Keeps "Observer Effect" under control

## **Derivative Metrics**



- Derivative Metrics can be automatically produced:
  - Counters => Rates
  - Gauges => Min, Mean, Max, Frequency distributions (histograms)

#### Metrics Rate Down-Sampling



- Metrics update rate is defined by the data handling rate:
  - E.g. rate of triggers for the ATLAS experiment is 100 kHz
- High update rates must be scaled down:
  - Take too much space in the data storage
    - 100 kHz of event rate => (8 + 8)\*3600\*10<sup>5</sup> = ~6 GB data per hour per single metrics
  - Cannot be visualized:
    - 4K displays have 3840 pixels along X axis
    - Can display data for 40ms only

#### Metrics Rate Down-Sampling



- Metrics values can be down-sampled by the API implementation:
  - Reduces recording rate
  - Simplifies storage requirements
- Output update interval can be made configurable:
  - A default value for all metrics
  - Individual values per specific metrics
- Transparent for the Applications and Communication components

#### Down-Sampling: Counters vs Gauges

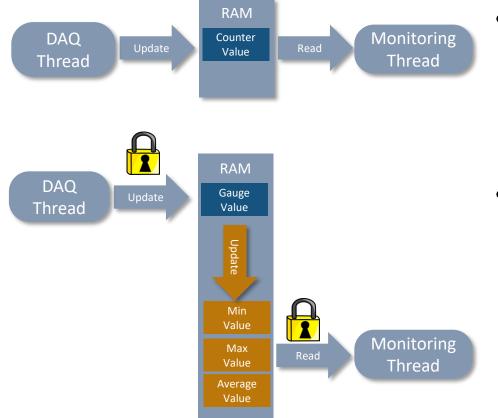
100 Counter: 90 80 Publish the last value for Occupancy (%) 70 each output update interval 60 50 • Gauge: 40 Publish three values for 30 20 each output update 10 interval: 0 Min, Average, Max • 200 0 100 300 Time (ms) Buffer Occupancy (%) Min Min Average Using Average only may hide 🗯 Max important information

#### The Observer Effect



- An observation affects the system:
  - It consumes resources (CPU, memory, network bandwidth)
  - It may affect performance of the monitored application
- Information must be passed to the Communication component <u>asynchronously</u>:
  - Monitoring information is updated by the <u>DAQ thread</u>
  - Down-sampling and publishing must be done by another thread
- Thread-safety must be considered:
  - But excessive thread-safety measures may hit the DAQ application performance

## Thread-safety Overhead



- Counters don't require a critical section:
  - Memory read/write operations on the modern Intel CPUs are atomic
- Gauge is different:
  - Monitoring Thread must not keep the lock when passing data to Communication component
  - Use a local copy

#### Thread-Safety Overhead

- Locking an unlocked mutex takes ~50 CPU cycles => less than 50ns:
  - If the mutex is locked this may lead to arbitrary delay
- Example: monitoring the buffer occupancy:
  - 10 kHz input rate:
    - Mutex locking takes 0.5ms every second => 0.05% overhead
  - 1 MHz input rate:
    - Mutex locking takes 50ms every second => 5% overhead

# Scaling up the Monitoring System



#### The HEP Experimental Realm

- A DAQ system of a modern HEP experiment consists of:
  - O(1K) computers and network devices
  - O(10K) SW applications
  - O(100K) Metrics

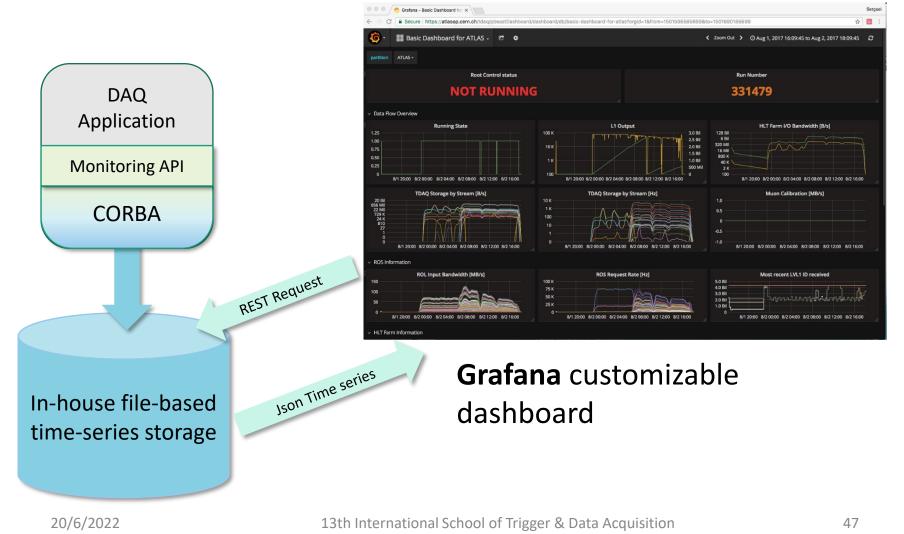


- A single gauge metrics for 24h run requires:
  - (8 + 8\*3)\*360\*24=**280**kB of RAM
- 100K Metrics => 28GB per day => 200GB per week => 10TB per year

#### Large Storage Implementations

- Traditional relational databases will not work well for largescale projects
- NoSQL distributed alternatives:
  - Whisper a lightweight, flat-file database format for storing timeseries data
  - InfluxDB a time-series database written in Go
  - **Cassandra** scalable, high availability storage platform
  - **MongoDB** a general purpose, document-based, distributed database

## The ATLAS Experiment: Web-based Metrics Monitoring

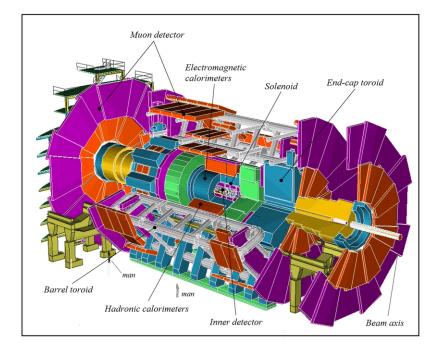




## DAQ Specialty: Data Quality Monitoring

#### How to Monitor the Detector?

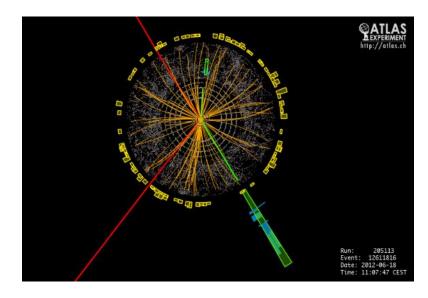
- Detectors of LHC experiments are incredibly complex devices:
  - Up to 10<sup>8</sup> output data channels
  - Mostly custom electronics
  - 40 MHz operational frequency
- Traditional monitoring would yield in O(1) PHz (petahertz) of metrics update rate:
  - These metrics are not even attempted to be produced
- However, DAQ system has a handle on these metrics...



#### **Detector Metrics**

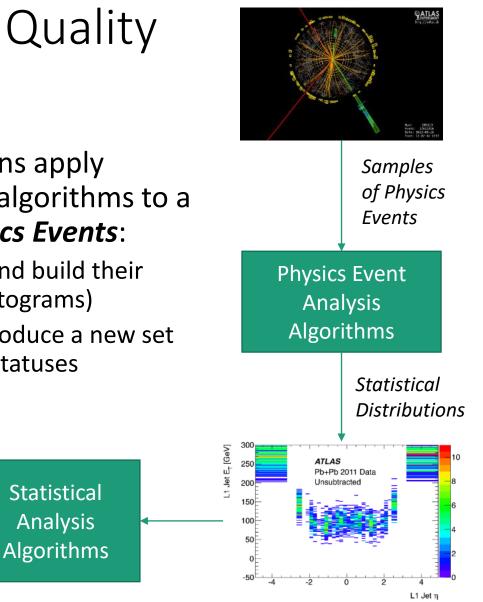
- Every Physics Event contains states of a sub-set of detector channels:
  - An expert can spot problems by looking into a graphical event representation
  - Such experts are not many and can't be in the Control Room 24/7

20489082 2057efb2	205a8616	2063cce2			20768ff7	99522077
	000000		000001	d04326b2	dd1234dd	0000002d
		0x61: MD		Run	0000009	03010000
g Header <sub>)aa8</sub>		arrel side		number	20128ec2	2017c212
dog <u>zo usoz</u> 2034	afb7400	(modulo 2	) 53c		J5829672	2063c2e2
207 5e2 2075d5b2	207aa892	a 207b	ed72ee7	00000000	00000000	00000002
3de510d4 dd1234dd	00000031	000 0009	04000,00	00610002	00000002	00000000
ee1234ee 00000009	03010000	00610002	00033dac	920117d5	00000aa8	00000081
2011ee42 efc22012	93222013	e2822014	97022017	e182201b	e0222025	eaa22027
84b22035 c5c2ccb2	2036ebc2	20389672	20508002	95a22051	d3172056	9ee22057
2060ad62 2061c4a2	2063ddb7	20649542	00000000	00000000	00000002	00000019
dd1234dd 00000029	00000009	01000000	00610003	00000002	00000000	<b>2011d80</b>
00000009 03010000	00610003	033dac	920117d5	00000aa8	00000081	000000
2031d692 20369542	2037ed92	409c92	ace22044	9a822046	a9e22047	3422048
e172205b c4872060	8f82206	data	c3f24000	00000000		2
aeaa0e15 dd1234dd	0000003	data ,	04000000	00610004	0000000	Trailer <sub>0</sub>
ee1234ee 00000009	03010000	00610004	00033dac	920117d5	00000aa8	00000081



## Automated Data Quality Analysis

- Dedicated DAQ applications apply standard physics analysis algorithms to a statistical sub-set of *Physics Events*:
  - Extract Detector Metrics and build their statistical distributions(histograms)
  - Analyze histograms and produce a new set of Metrics – Data Quality statuses



## Summary: The Key Points



- Have your Monitoring System API ready from the beginning of the main project
- ✓ Use standard Monitoring APIs whenever it is possible:
  - e.g. Logging API
- ✓ Think carefully when designing a custom API:
  - It must not depend on a particular technology
- ✓ The Monitoring System implementation may evolve during DAQ system development
- ✓ Use existing solutions for Communication and Visualization components:
  - In-house development must be well justified