

# Data Analytics – Use Cases, Platforms, Services

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# Analytics and Big Data Pipelines – Use Cases

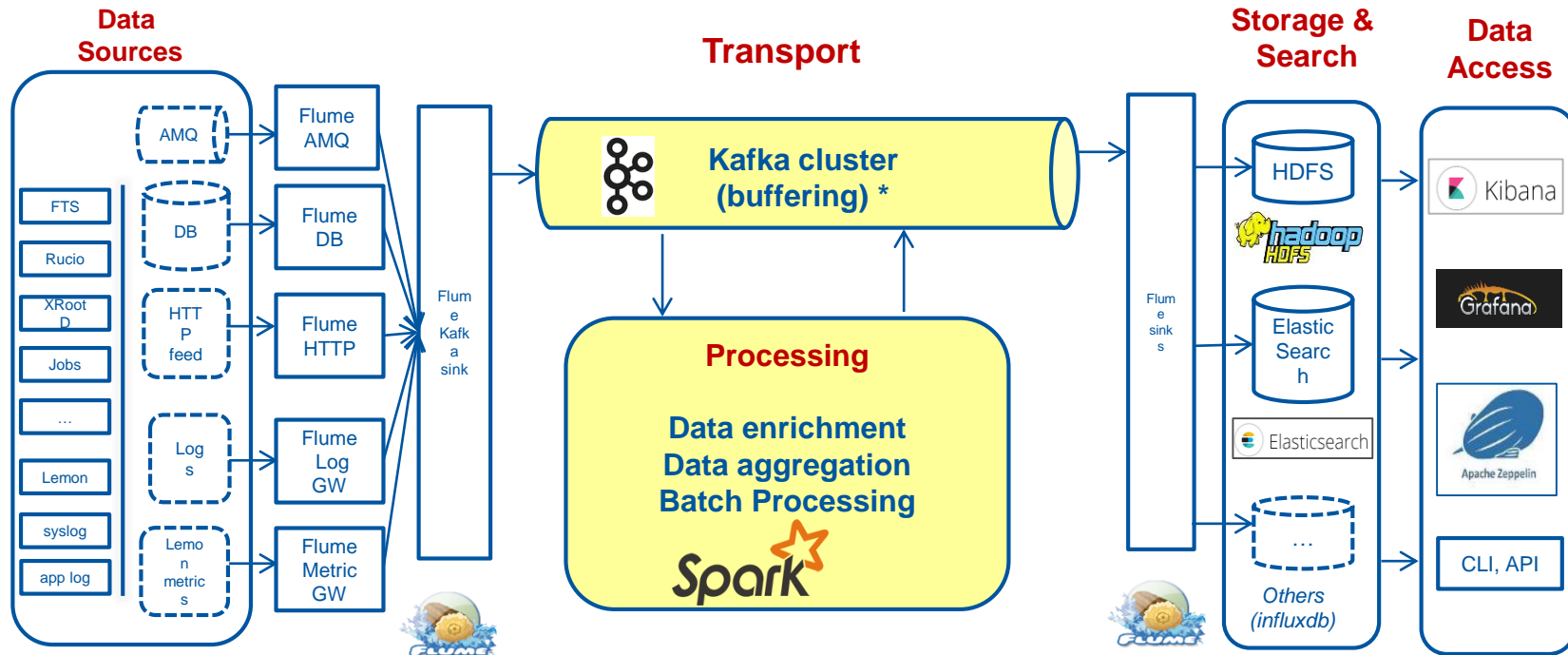
- Many use cases at CERN for analytics
  - Data analysis, dashboards, plots, joining and aggregating multiple data, libraries for specialized processing, machine learning, ...
- Communities
  - **Physics:**
    - Analytics on computing data (e.g. studies of popularity, grid jobs, etc) (CMS, ATLAS)
    - Development of new ways to process **ROOT** data, e.g.: data reduction and analysis with spark-ROOT by CMS Bigdata project, also TOTEM working on this
  - **IT:**
    - Analytics on IT monitoring data
    - Computer security
  - **BE:**
    - NX CALS – next generation accelerator logging platform
    - BE controls data and analytics
  - More:
    - Many tools provided in our platforms are popular and readily available, likely to attract **new** projects
    - E.g. Starting investigations on data pipelines for IoT (Internet of Things)

# “Big Data”: Not Only Analytics

- Data **analytics** is a key use case for the platforms
- Scalable workloads and parallel computing
  - Example work on data reduction (CMS Big Data project) and parallel processing of ROOT data (EP-SFT)
- **Database**-type workload also important
  - Use Big Data tools instead of RDBMS
  - Examples: NXCALS, ATLAS EventIndex, explorations on WINCC/PVSS next generation
- Data pipelines and **streaming**
  - See example of monitoring and Computer security (Kafka development with help of CM)
  - Also current investigations on IoT (project with CS)

# New IT Monitoring

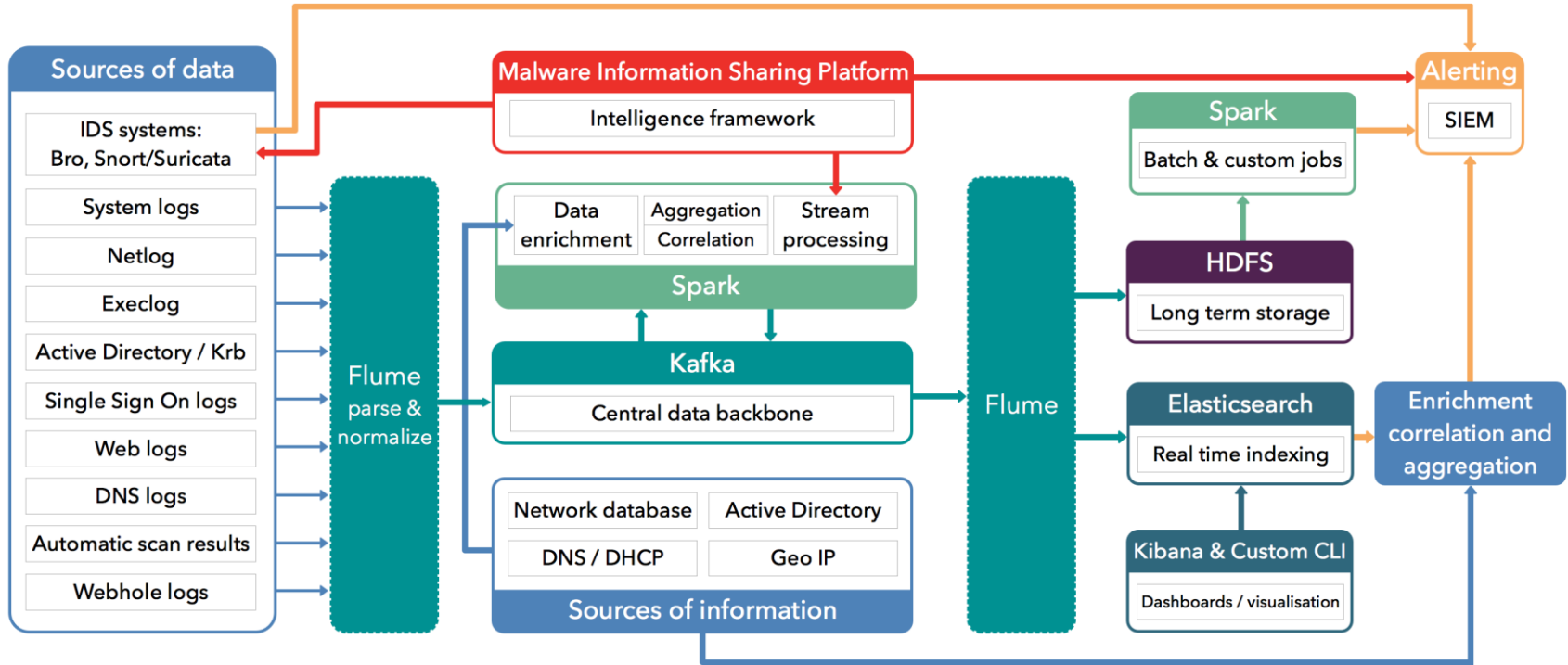
Critical for CC operations and WLCG



- Data now 200 GB/day, 200M events/day
- At scale 500 GB/day
- Proved effective in several occasions

# Computer Security

## intrusion detection use cases



# Example, BE-NXCALS

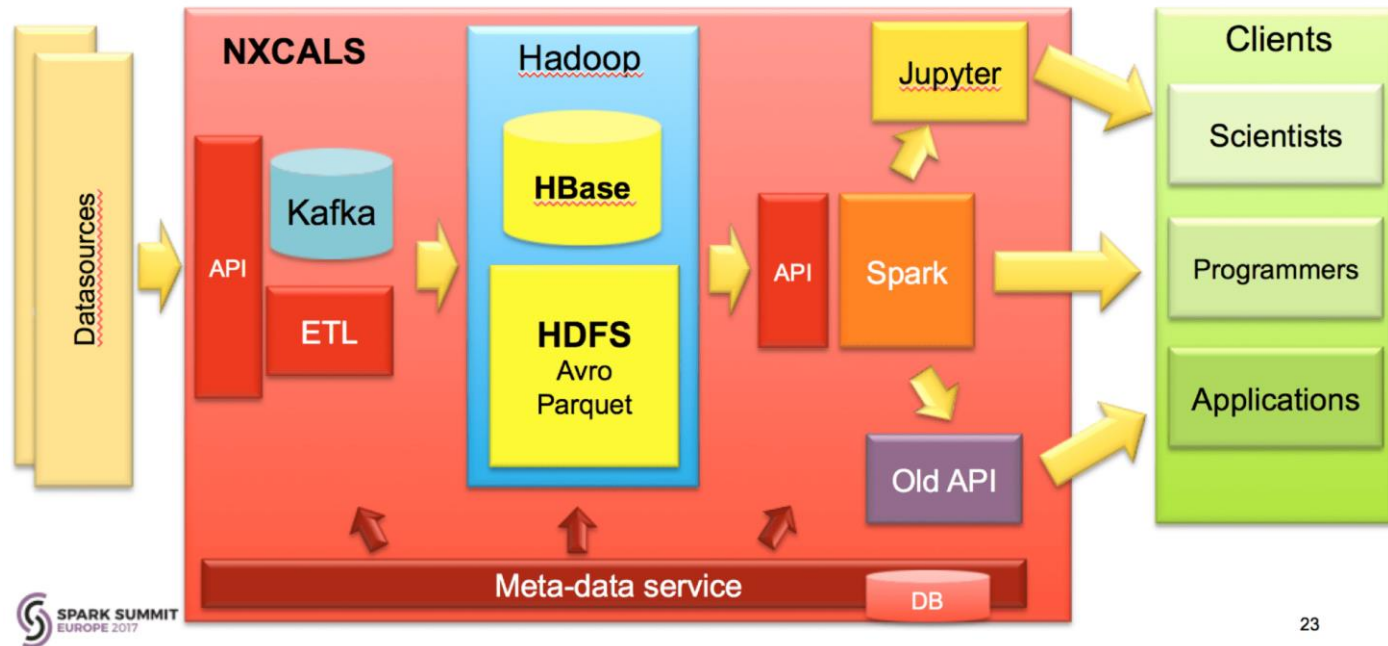
Many parts

- Streaming
- Online system
- APIs for data extraction and analytics

Note:

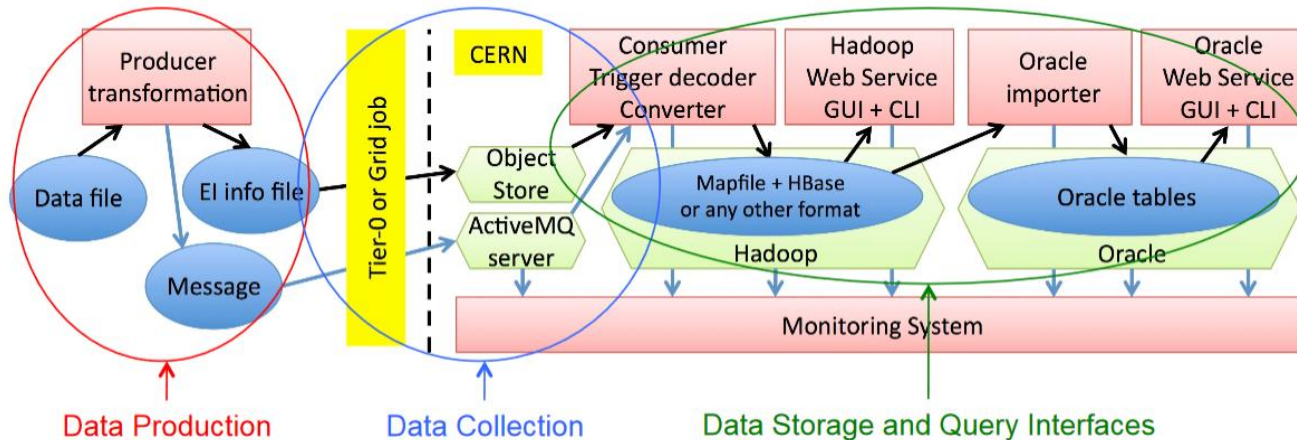
This will replace and enhance current platform based on RDBMS

## NXCALS Architecture



# ATLAS EventIndex

- Searchable catalog of ATLAS events
  - First “Big Data” project in our systems
  - Over 80 billions of records, 140TB of data



Credits: Dario Barberis, 2017

# Goal: Meet Use Cases With Managed **Services** for Data Engineering and Analytics

- **Platform**
  - Capacity planning and configuration
  - Define, configure and support components
  - Use tools and ideas from “Big Data”/ open source communities
- Running central **services**
  - Build a team with domain expertise
  - Share experience, provide consultancy
  - Economy of scale



# Highlights of “Big Data” Components

- Apache Hadoop clusters with YARN and HDFS
  - Also HBase, Impala, Hive,...
- Apache Spark for analytics
  - Apache Kafka for streaming
- Data: Parquet, JSON, ROOT
- UI: Notebooks/ SWAN



# Tools and Platforms – Ecosystem is Large

- Many tools available: from HEP + other sciences + open source
  - **Challenges** and **opportunities**: build your solution + expect natural evolution
  - Components are modular, **users** adoption and traction drives efforts

- Front ends



- Engines and Systems



- Storage and data sources



Databases



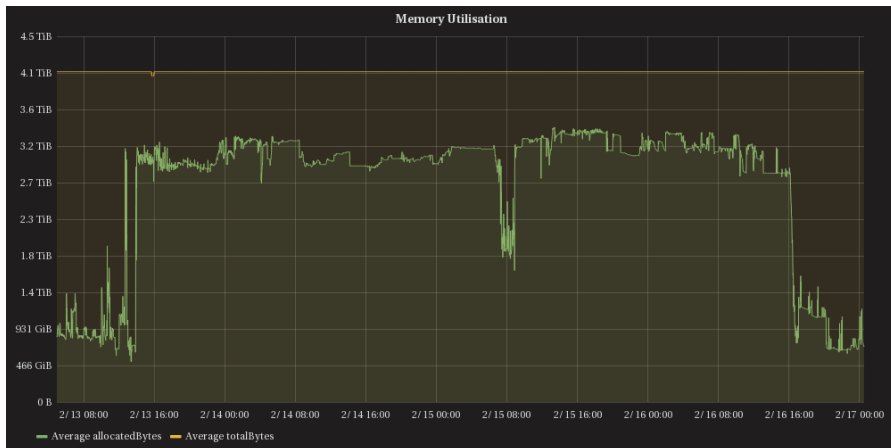
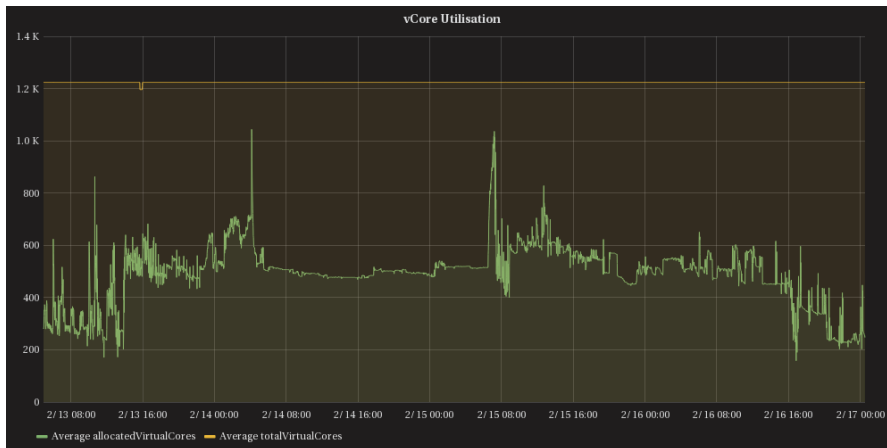
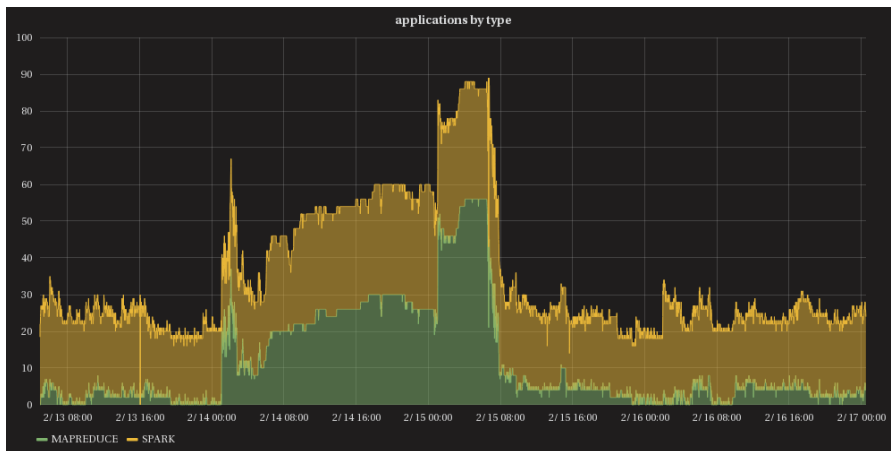
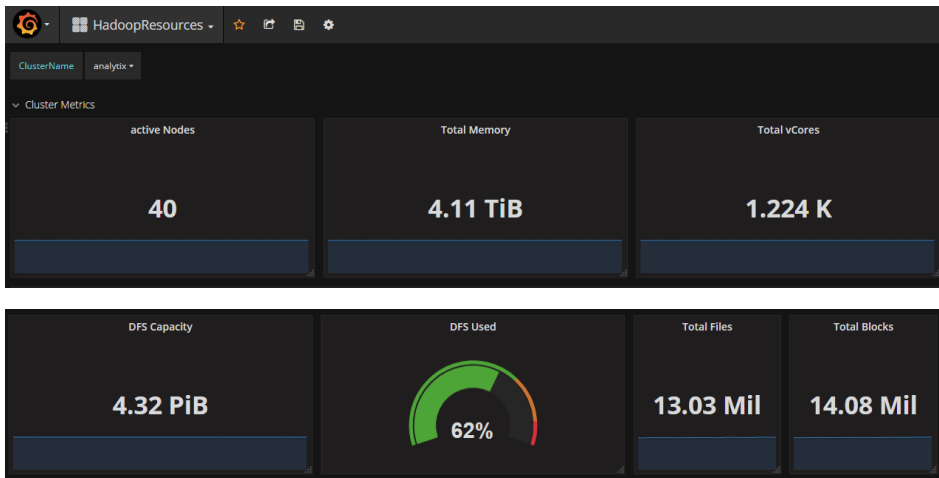
- Data formats: ROOT, Parquet, JSON,...

# Hadoop Clusters at CERN IT

- 3 current production clusters
  - + environments for NXCALS DEV and HadoopQA
- A new system for **BE NXCALS** (accelerator logging) platform
  - Currently being commissioned

Cluster Name	Specification	Primary Usage
lxhadoop	18 nodes (cores - 520,Mem - 701GB,Storage - 1.29 PB)	ATLAS EventIndex
analytix	40 nodes (cores - 1224,Mem - 4.11TB,Storage - 4.32 PB)	General Purpose
hadalytic	12 nodes (cores - 364,Mem - 580GB,Storage - 2.081 PB)	Development Cluster (old BE prod)
NxCALS	24 nodes (cores - 1152,Mem - 12TB,Storage - 4.6PB,SSD - 92 TB)	Accelerator Logging Service

# Workload Metrics



# Projects and Activities

- **Spark** integration with **SWAN** (Jupyter notebook service)
- Spark for **HEP** (CMS Big Data, Spark and EOS)
- **Kafka** Service and Self-service tools
- Kudu: online and analytics, best of HBase and HDFS?
- Spark as a service (Spark and Kubernetes)
- Courses, **consultancy**, forums, workshops and conferences, documentation, etc

# Spark for Physics

- Promising results
  - See ACAT 2017: “CMS Analysis and Data Reduction with Apache Spark”
  - Processing ROOT data using Big Data tools (Spark)
  - Aligned with the goals of the HEP software foundation activities
- Currently working with CMS Bigdata and Intel on openlab
  - Developed Hadoop-XRootD connector (read EOS from Spark)
  - Working on large-scale reduction and analysis
- Also collaborating with Totem

# Spark as a Service and Cloud

- Exploring Spark on Kubernetes
  - Separate compute cluster from storage
  - Already used in industry, now mainstream with latest Spark
  - **Exploration** phase for us at this stage
- Motivations
  - Can this help to **scale** out on large workloads and help streamline **resource** allocations?
- Context
  - For the use case of using Spark to process physics data on EOS, running Spark on Hadoop with HDFS has limited advantages
  - Spark is basically a library and needs a cluster manager: YARN/Hadoop, Mesos, Stand-alone, since Q1 2018 also Kubernetes

# Spark Integration with SWAN – Towards an Analytics Platform



- Goal: become go-to platform for analytics at CERN
  - Lower the barrier to start doing **analytics** and provide a pre-configured environment with the most used **tools**
- Integration of **SWAN** with **Spark** clusters in IT
  - Key component for BE-NXCALS: chosen as user interface
  - Already used by several pilot users too, for example for monitoring data analytics
  - opening to all users later in 2018. Currently SWAN default for Spark is to use “local mode” (no cluster)
  - Joint efforts with EP-SFT, IT-ST, IT-CM on integrating functionality, security and monitoring aspects



# Jupyter Notebooks and Analytics Platforms

## Apache Spark

### LHCb Opendata

This notebook illustrates a nice interplay between LHCb Opendata, Spark and ROOT - all teaching interesting flavour Physics!

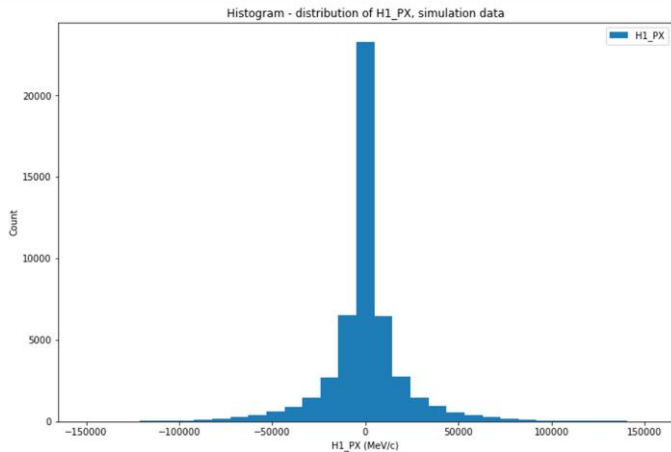
#### Plotting a feature:

You can plot any feature of the data in a histogram. Choose any suitable binning that allows you to observed the distribution of the variable clearly. In this example, we will plot a histogram for the first kaon candidate's momentum x-component (H1\_PX):

```
In [8]: # Plot a histogram of the distribution of the H1_PX variable, using Pandas

# This is a basic solution that moves all the data from the Spark DataFrame
# into a Python Pandas DataFrame. It's OK for small data sets, but it has scalability issues

h1px_data = sim_data_df.select("H1_PX").toPandas() # select H1_PX data and moves it to Pandas
h1px_data.plot.hist(bins=31, range=[-150000, 150000], title="Histogram - distribution of H1_PX, simulation data")
xlabel('H1_PX (MeV/c)')
ylabel('Count');
```



## SWAN Customisation

Specify the parameters that will be used to contextualise the container which is created for you. See the [online SWAN guide](#) for more details.

#### Software stack [more...](#)

89

#### Platform [more...](#)

x86\_64-slc6-gcc49-opt

#### Environment script [more...](#)

e.g. \$CERNBOX\_HOME/MySWAN/myscript.sh

#### Number of cores [more...](#)

Number of cores to associate to the container.

1

#### Spark cluster [more...](#)

Hadalytic

Start my Session

# Jupyter Notebooks and Analytics Platforms



Streaming SQL ML Graph



ROOT  
Data Analysis Framework



# Bringing Database Workloads to Big Data

- Goal: **migrations** from RDBM to “Big Data”
  - Technically **challenging**: need systems that can perform “Hybrid transactional/analytical processing” (HTAP)
  - Not all workloads can fit, but when applicable can provide gain in performance and **reduced cost**
- Technology exploration:
  - **Apache Kudu**: a storage system that can be used with Big Data tools, with Indexes and columnar features
- Testing: next gen ATLAS Event Index, next gen WINCC (PVSS) archiver

# Users and community

- Develop value of the services for users and community
  - We have developed and delivered short **courses** on Spark and Hadoop
    - More session(s) scheduled in 2018 + planning course on Kafka
  - Plannig -> presentations at relevant forums (ASDF, IT Tech, ITUM)
  - Consultancy and projects with relevant communities
- Linking with communities outside CERN
  - Contacts with **industry** and open source communities
    - (Intel openlab, conferences, etc)
  - Others in HEP: Already contacted selected **T1s** and other centers (Sara, BNL, Princeton) in 2017 for sharing ideas on Hadoop config + CMS Bigdata (Fermi Lab, Princeton, Padova) + TOTEM (team in Warsaw)
  - Potential: contacts with "other sciences" (astrophysics, biology/medical)

# Challenges

- Platforms
  - Provide evolution for **HW** – currently plan to do “rolling upgrades” of ~ 25% of the platform yearly, some of the current HW is quite old
- Service
  - Build robust service for **critical** platform (NXCALS and more) using custom-integrated **open source** software solutions in constant **evolution**
  - Support production services (IT monitoring, Security, EventIndex)
  - Evolve service configuration and **procedures** to fulfil users needs
  - Further grow value for community and projects
- Knowledge and experience
  - Technology keeps evolving, need to learn and adapt quickly to **change**

# Conclusions

- Hadoop, Spark, Kafka services
  - Unified support, engineering and consulting
  - Started work for specific use cases, experience led to opening central services to more communities (Hadoop and Spark, Spark on SWAN)
- Production service support
  - BE-NXCALS, critical service starting in 2018
  - Support for IT-Monitoring, Security and ATLAS EventIndex
  - Still developing and streamlining services (configuration, monitoring, backup, tools, etc)
- Growing value for user communities
  - Developing projects
    - Openlab with CMS Big Data and Intel for Physics analysis and data reduction with Spark
    - Evolution of WINCC/PVSS and Atlas EventIndex
  - Leverage experience and bring solutions around Big Data to a growing community
    - Lower barrier to entry (Spark on notebook), organize courses, presentations, and knowledge sharing activities in general