# Analysis in LHC RUN3 (Alice case)

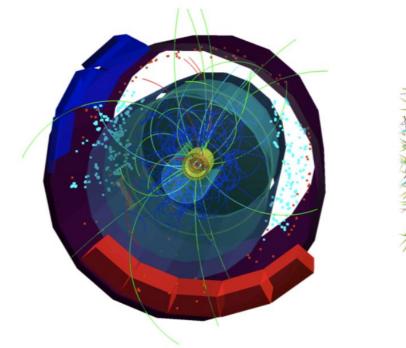
M. Al-Turany, GSI/IT

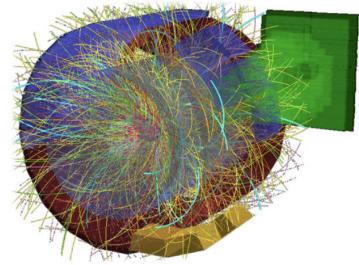
Thanks to:

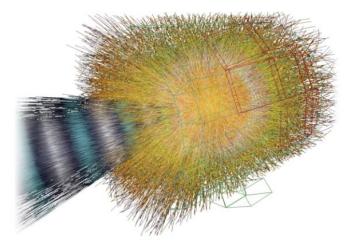
#### Giulio Eulisse, Peter Hristov, Ruben Shahoyan, Thorsten Kollegger, Killian Schwarz

#### Alice in RUN2 about O(1) kHz single events more than 4 Gbytes/s to Storage









02.11.19

Pb - Pb

#### Analysis in RUN2:



- Organized analysis
- Event-oriented data model: trees of ESD & AOD/delta AOD, but also kinematics, ESD friends, track references, tags
  - Access to the different data via handlers
- Possibility to run in local, Proof, GRID, event mixing modes
  - Services: I/O, event loop, merging of results, bookkeeping
  - LEGO trains
- All user code on GitHub (alisw/AliPhysics) and built centrally on CVMFS

### Analysis Trains:

Analysis tasks organized in trains (dependencies, I/O):

- Read data once,
- process many times,
- benefit from common processing



## ALICE Analysis Facilities (Run1/Run2)

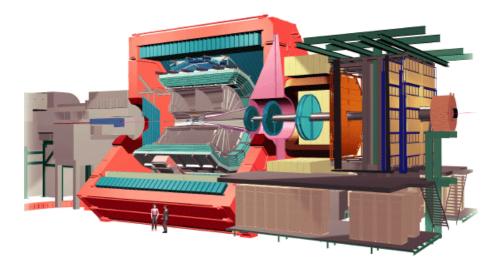
- PROOF-based facilities: CERN, Lyon, GSI, Torino and many other places
  - Local data sets
  - Running on native resources
  - Using shared file system
  - Remote access from laptop/desktop

## ALICE Upgrade

- The Inner Tracking System (ITS) will be replaced with a new, highresolution, low-material detector
- The Time Projection Chamber (TPC) will be upgraded with replacement of the chambers by Gas Electron Multipliers (GEMs) and a new pipelined readout electronics based on a continuous read-out scheme
- The forward trigger detectors and the electronics of the Transition Radiation Detector (TRD), the Time Of Flight (TOF), and several other detectors will be upgraded

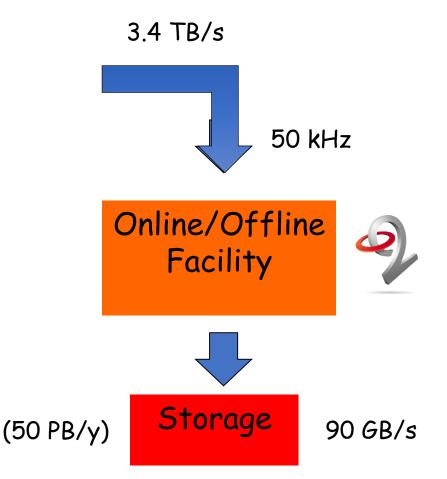


#### ALICE Upgrade



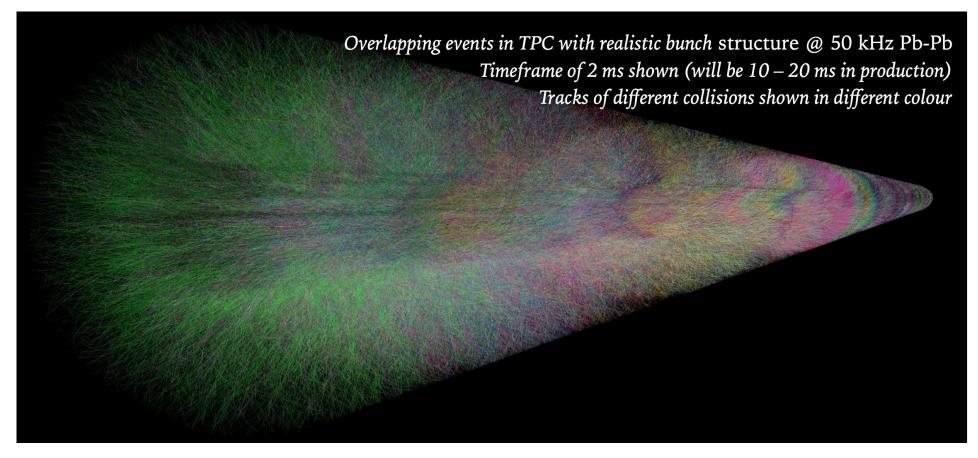
- Aim is to reduce data volume by doing (quasi) online reconstruction
  - Each and every event needs to be processed, no rejection
- High Throughput (and not Performance) Computing problem

- continuous readout
- x50 event rate

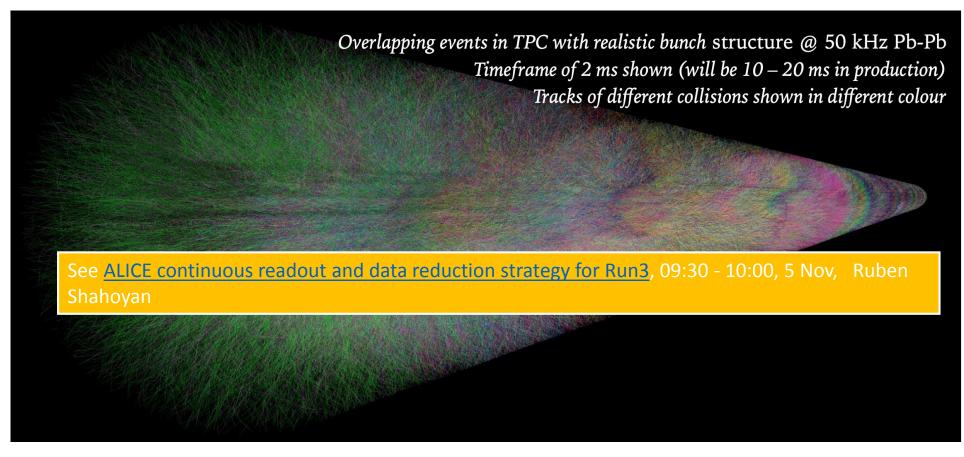


#### Online Reconstruction: O2 Facility CR0 or Computing Computing Room 1 Room 0 Meyrin Computing CR 1 CR 0 Center Detectors Read-out FLP to EPN Storage 9000 Network Network Read-out CTF: 35 GB/s Links DS С Tier 0 - - D-EPN R FLP U CTF:5-20 GB/s Tiers 1 \* - - De AOD:5-20 GB/s Analysis 14 A 🖗 3.4 500 Write: 100+20 GB/s Facilities TB/s Read: under review GB/s \* Baseline correction, Data volume reduction Data by online tracking, zero suppression Storage cluster finder. Asynchronous processing

### Alice in RUN3 50 kHz of continuous readout data. 90 Gbytes/s to Storage (50 PB/y)



### Alice in RUN3 50 kHz of continuous readout data. 90 Gbytes/s to Storage (50 PB/y)



# Compared to RUN2

- Reconstruct 50x more events online
- Store 50x more events
- continuous readout (TPC data ) in combination with data coming from triggered detectors.

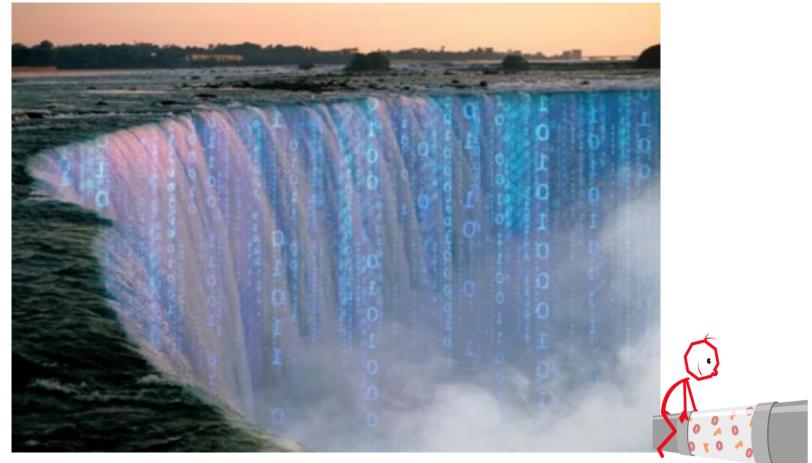


# What to do?

ALICE can cope with the challenges of Run3 only by a radical redesign of its software and computing architecture.



### A data-flow based model:



#### Message Queues based multi-processing

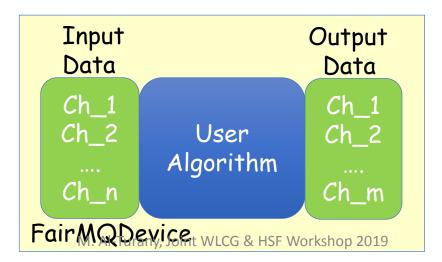
### A data-flow based model:

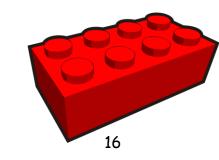


#### Message Queues based multi-processing

# ALFA building block (FairMQ Devices)

- Message Queues for input/output
- Device takes/passes ownership of data
- Framework user sees only the callback to his algorithm
- Different channels can use different transport engines



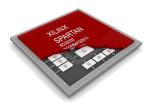


# Message format?





FairMQ does not impose any format on messages.







It supports different serialization standards

- BOOST C++ serialization
- Google's protocol buffers
- ROOT
- Flatbuffers
- MessagePack
- User defined

s boost

Protobuf Protocol Buffers

MP



S

lava





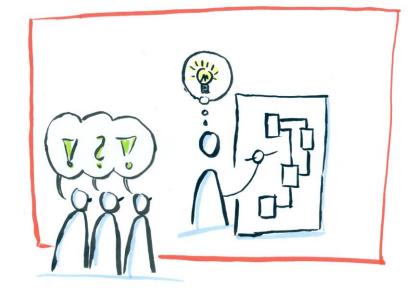
### Software framework: Transport Layer



- Uses FairMQ message passing toolkit (GSI development)
- Abstracts the network fabric
- Defines the core building blocks in terms of devices
- Implements the communication between them

<u>ALFA: A framework for building distributed applications</u> Track 5 - Software Development 11:30 - 11:45

### Software framework: O2 Data Model



- ALICE-specific description of the messages between devices
- Computer language agnostic, extensible, efficient mapping of the data objects in shared memory or to the GPU memory
- Supports multiple data formats and serialization methods

### Software framework: Data Processing Layer

Data Analysis using ALICE Run3 Framework, S. Eulisse, 16, 11:45

- Simplifies the life of the end user
- Allows to describe computation as a set of data processors implicitly organized in a logical data flow transformation
- A defined data flow is run by a single executable the DPL driver
- Includes a powerful GUI for logs/metrics and debugging
  - Especially helpful for individual users

# Analysis in RUN3:



Problem:

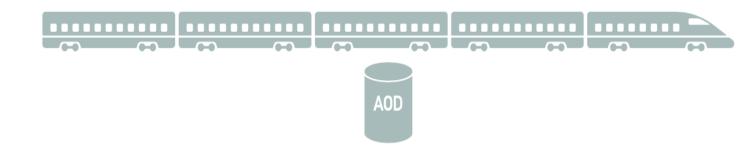
Analysis remains I/O bound in spite of attempts to make it more efficient by using the train approach

Data Analysis using ALICE Run3 Framework, 5 Nov, G.Eulisse, T6, 11:45

# Analysis in RUN3: (Solution)



- Retain concepts that worked: analysis trains, centralized code, abstraction framework
- Use better compression algorithms
- **Recompute** quantities on the fly rather than storing them.
- Flat data structures
- Only AODs for analysis



# Requirements for the AOD format

- AOD's data format will have to play well with AliceO2 message passing, shared memory backed, distributed nature.
- Zero-{Copy, Serialisation, Adjustments}:
  - we want to be able to reuse data between processes.
- Growable: ability to extend columns on the fly.
- **Prunable**: ability to drop columns on the fly.
- Skimmable: ability to select only certain rows.
- Strategy: we are willing to lose some degree of generality for performance.

# Apache Arrow

- Data Analysis using ALICE Run3 Framework, • Apache Arrow as backing store for the message passing.
- Arrow fits well to represent column oriented data, while providing some level of flexibility for nested data via the usual record shredding.
- Using Apache Arrow allows for seamless integration with a larger ecosystem of tools, like Pandas or Tensorflow.

# ALICE Analysis Facilities (Run3)

Motivation



- Analysis is the least efficient of all workloads that we run on the Grid
- I/O bound in spite of attempts to make it more efficient by using the analysis trains
- Solution
  - Collect AODs on a few dedicated sites that are capable of locally processing quickly large data volume
  - Typically (a fraction of) HPC facility (20-30'000 cores) and 5-10 PB of disk on very performant file system
  - Run organized analysis on local data like we do today on the Grid

# Analysis facility @ GSI (Prototype for RUN3)

- GSI Tier 2 Resources
- Full AOD set 2015 Pb-Pb (LHC150, about 250 TB)
- Shared file system (Lustre) + xrootd client plugin (0.6PB)
- Performance tests suggest that the target throughput rate of 10 PB/day can be achieved

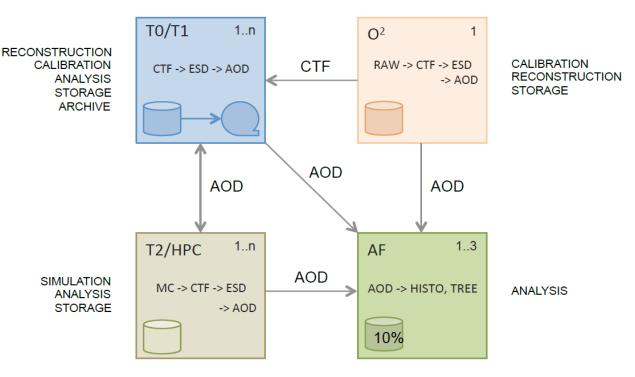
A prototype for the ALICE Analysis Facility at GSI (2018) https://indico.cern.ch/event/587955/contributions/2937941/

### Computing model in a single figure

Grid Tiers will be mostly specialized for given role

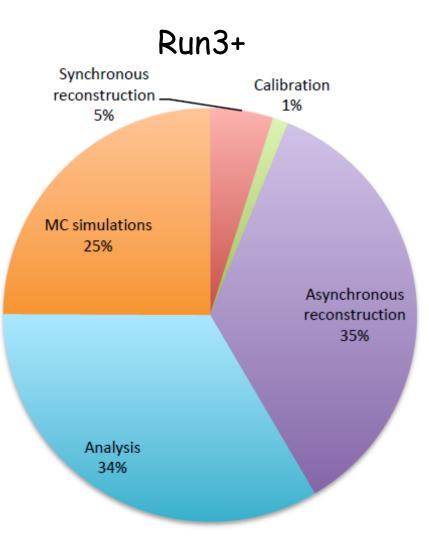
- O2 facility (2/3 of reconstruction and calibration),
- T1s (1/3 of reconstruction and calibration, archiving to tape),
- T2s (simulation)
- All AODs will be collected on the specialized Analysis

Facilities (AF) capable of processing ~5 PB of data within  $\frac{1}{2}$  day timescale The goal is to minimize data movement and optimize processing efficiency

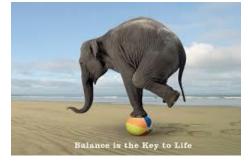


### Resources share projection

Run2 Reconstruction 14% Analysis 14% MC simulations 72%



### Summary



- Message Queues based solution (microservices) as a new paradigm for ALICE software
  - Different topologies of tasks can be adapted to the problem itself, and the hardware capabilities
- Apache Arrow as in memory backing store simplifies the interoperability with a number of OpenSource tools.
- Performance tests of the proto type AF at GSI, suggest that the target throughput rate of 10 PB/day can be achieved