

Summary of TDAQ Sessions

CPAD Instrumentation Frontier Workshop 2021

Georgia Karagiorgi, Columbia University
and
Isobel Ojalvo, Princeton University

March 22, 2021

Exciting developments in Trigger and DAQ Instrumentation

Day 1

12:00	Mu2e TDAQ and slow control systems <i>Stony Brook, NY</i>	<i>Antonio Gioiosa</i> 12:00 - 12:15
	Development of the Mu2e electromagnetic calorimeter front-end and readout electronics <i>Stony Brook, NY</i>	<i>Franco Spinella et al.</i> 12:20 - 12:35
	Does anybody really know the time it is? <i>Stony Brook, NY</i>	<i>Roger Rusack</i> 12:40 - 12:55
13:00	FELIX: the Detector Interface for the ATLAS Experiment at CERN <i>Stony Brook, NY</i>	<i>Alexander Paramonov</i> 13:00 - 13:15
	The Expandable Modular ATCA hardware design for high-energy physics experiment applications <i>Stony Brook, NY</i>	<i>Alexander Madorsky</i> 13:20 - 13:35
	Break <i>Stony Brook, NY</i> 13:40 - 14:00	
14:00	Liquid Argon Time Projection Chamber Trigger Development with MicroBooNE and SBND <i>Stony Brook, NY</i>	<i>Daisy Kalra et al.</i> 14:00 - 14:15
	Trigger and Data Acquisition for the Mu2e-II experiment <i>Stony Brook, NY</i>	<i>Richard Bonventre et al.</i> 14:20 - 14:35
	Streaming data acquisition system for CLAS12 Forward Tagger <i>Stony Brook, NY</i>	<i>Mariangela Bondi</i> 14:40 - 14:55
15:00	Designing a 30 MHz GPU trigger, the LHCb experience <i>Stony Brook, NY</i>	<i>Vladimir Gligorov</i> 15:00 - 15:15
	Detector design for a Muon Collider experiment <i>Stony Brook, NY</i>	<i>Sergo Jindariani</i> 15:20 - 15:35

Day 2

12:00	Real-time analysis in Run 3 with the LHCb experiment <i>Stony Brook, NY</i>	<i>Mika Vesterinen</i> 12:00 - 12:15
	Towards an Interpretable Data-driven Trigger System for High-Throughput Physics Facilities <i>Stony Brook, NY</i>	<i>David Miller</i> 12:20 - 12:35
	Global Trigger for the ATLAS Phase-II upgrade <i>Stony Brook, NY</i>	<i>Jochen Heinrich</i> 12:40 - 12:55
13:00	The Challenges of Machine Learning at the Edge <i>Stony Brook, NY</i>	<i>Audrey Corbell Therrien</i> 13:00 - 13:15
	Coprocessors as a service for accelerated inference of DL algorithms <i>Stony Brook, NY</i>	<i>Jeffrey Krupa</i> 13:20 - 13:35
	Break <i>Stony Brook, NY</i> 13:40 - 14:00	
14:00	Triggering on Long-Lived Particles decaying to Hadronic Showers in CMS Muon System <i>Stony Brook, NY</i>	<i>Ka hei martin Kwok</i> 14:00 - 14:15
	Real-time Artificial Intelligence for Accelerator Control: A Study at the Fermilab Booster <i>Stony Brook, NY</i>	<i>Christian Herwig</i> 14:20 - 14:35
	hls4ml: enabling real-time deep learning in HEP trigger and DAQ systems <i>Stony Brook, NY</i>	<i>Jennifer Ngadiuba et al.</i> 14:40 - 14:55

A total of 18 contributions spanning Mu2e, ATLAS, MicroBooNE, SBND, LHCb, CMS, related experiment-specific upgrades/developments, and experiment-agnostic developments and applications

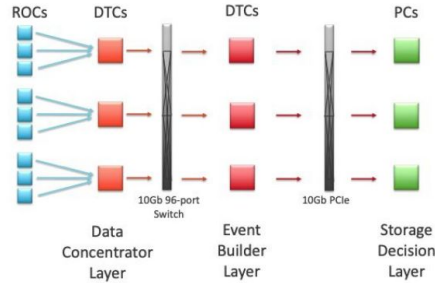
Common TDAQ Themes in 2021

- Next-generation TDAQ Systems developed as scalable systems with high level of parallelization, for high-throughput facilities, with an eye toward “self-driving” systems (see, e.g., Pie in the Sky talk by D. Miller)
- New TDAQ designs and performance are benefiting from **AI developments**, **hardware processor advancements**, and **design tools developments**
- High degree of cross-experiment collaboration and co-development

Highlights: Mu2e TDAQ and slow control systems

Mu2e requirement is to process 200k events/s

Readout scheme designed with capability of 40 GB/s data readout → 280 MB/s to disk

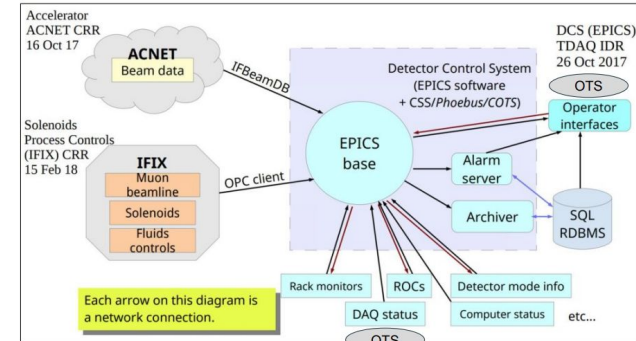
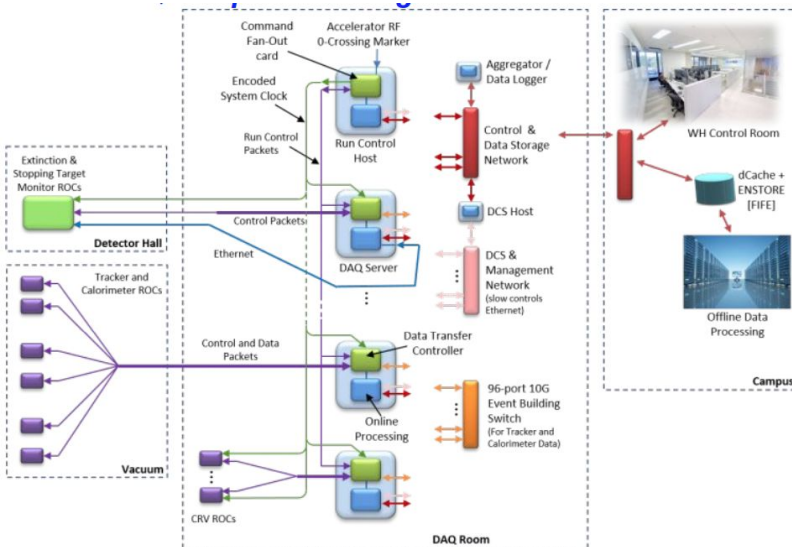


- Mu2e Experiment is under construction at Fermilab and will be ready for data taking next years
- Mu2e TDAQ and slow control are in large part developed according to the requirements (200K events/s for data taking) and hardware tests are going on
- Slow control integration in the online DAQ system, *otsdaq*, provides an advanced slow controls monitoring, an interface to send *otsdaq* front-end DAQ hardware, data processing and DQM slow controls informations to **EPICS**, and a real configuration and Integration with the *otsdaq* State Machine

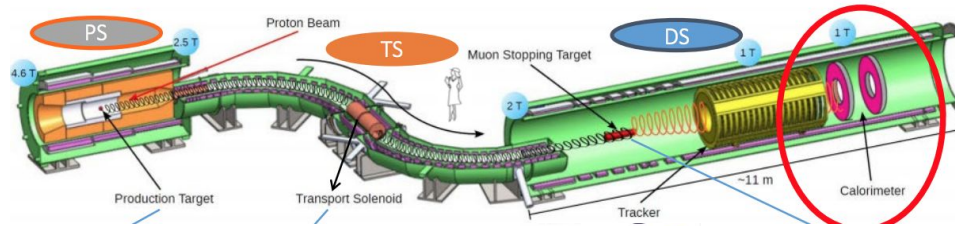
Online DAQ is off-the-shelf DAQ (otsdaq) based on artdaq



Provides a library of supported front-end boards and firmware modules, and integrated Run Control GUI and readout software



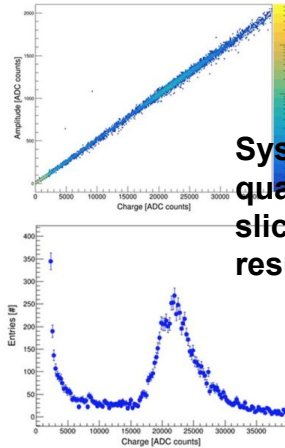
Highlights: Development of the Mu2E electromagnetic calorimeter front-end and readout electronics



DIRAC: custom waveform digitizer

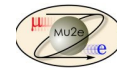


Intense campaign of tests to determine components



System design qualified with slice test results

Fig.B: Energy distribution deposited by cosmic rays (ADC counts)



Calorimeter electronics scheme



Disks x2

Crate x10

FEE x10 / board
(MPPC x2 / FEE)

CsI Crystal

Custom FEE

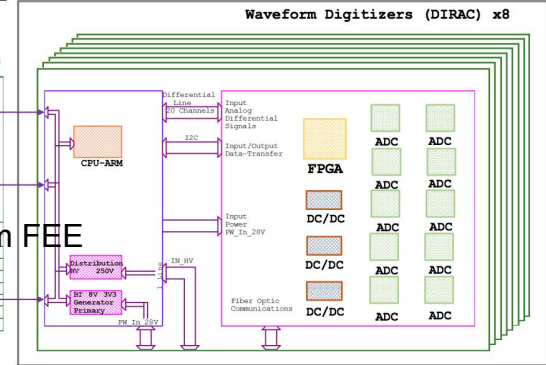
@ 105 MeV

Calorimeter requirements

- energy resolution $\sigma_E/E < 10\%$
- timing resolution $\sigma(t) < 200$ ps
- position resolution < 1 cm
- Work in vacuum @ 10^{-4} Torr
- 1 T Magnetic Field

CsI crystals + SiPMs

High radiation environment



- Sampling frequency of **200 MHz**
- ADC with **12 bits resolution**

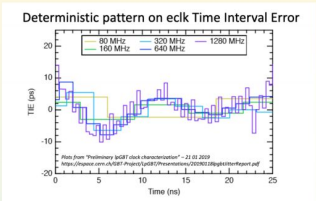
Highlights: “Does Anybody Really Know What Time it is?”

State-of-the-Art Today

LpGBT used to distribute a high precision clock derived by clock-recovery from the 2.56 Gbs downlink control signal.

- LpGBT-v0
- Random jitter 2.2 ps
 - Deterministic jitter peak-to-peak 25 ps.

Source identified and LpGBT-v1 expected to reduce deterministic jitter.



From Talk by T. Kugathasan December 2020.

As we push the limits of time measurements to ~1 ps, we need to have reference clocks that are stable at < 1ps.

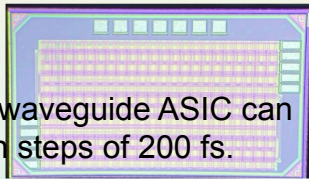
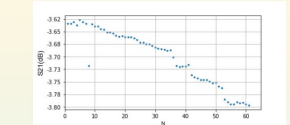
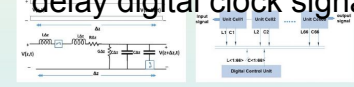
How Do We Correct For Drifts?

To solve the problem of how to align clocks that drift we have made a multi-cell planar waveguide in TSMC 65 nm process.

Digitally Controlled Phase Shifter — DCPS

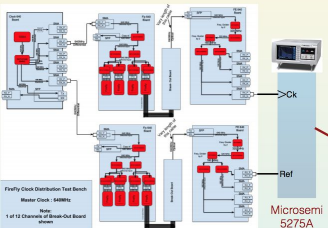


Digitally controlled planar waveguide ASIC can delay digital clock signal in steps of 200 fs.



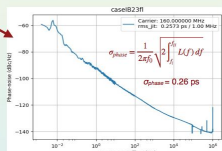
Measured delay step is 200 fs

‘Pure’ Clock Distribution System



Microsemi 5275A

RF Quality Fanout from ON Semiconductors. NB7VQ1005MMING — max data rate 10 Gbps.



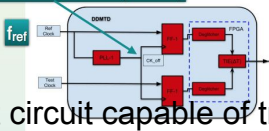
Phase Noise Plot 0.1 Hz - 1 MHz

Use DDMTD to measure time drifts averaged over many cycles.

Basic method that goes back to FM is radio is to heterodyne the signal.

Digital Dual Mean Time Difference (DDMTD) circuit*

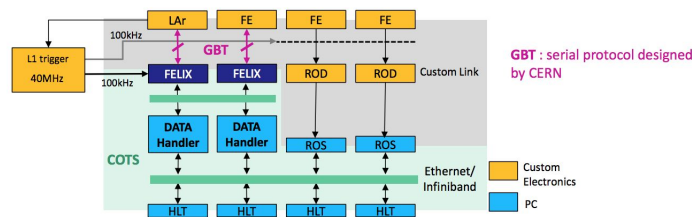
Offset clock with $f_{off} = f_{ref}(1 - 1/N)$



Low-cost circuit capable of tracking clock drifts at sub-ps level

Pure clock distribution system demonstrates sub-ps jitter levels, exceeding current state-of-the-art with clock recovery.
→ Can deliver stable clock with low jitter and low wander.

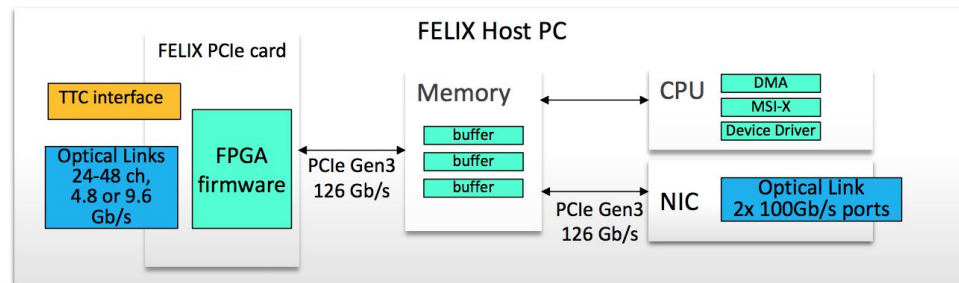
Highlights: **FELIX**: The new detector readout system for the **ATLAS** experiment



FELIX
Front End Link eXchange

- FELIX is a router between custom serial links and a commodity network
 - Takes advantage of the latest technology to simplify the ATLAS readout
- In LHC Run-3 (2021-2023) FELIX will be used for selected detectors and trigger systems.
 - FELIX firmware and the software are mature
 - Most of the boards have been produced
- Ongoing efforts:
 - Integration with the ATLAS on-detector (front-end) systems
 - Development of FELIX board and firmware for LHC Run-4.

FELIX hardware platform



PCIe card with FPGA chip + Host PC + NIC

Highlights: X2O: A Modular ATCA Design with Cost Optimization Options

A modular system in ATCA standard, designed with cost optimization in mind, and customizable for user's needs

Approach: eliminate base board, use large heat sink as mechanical platform,

Several modules: FPGA modules attached to heat sink, optical module near standard front panel, power module near backplane

Cost range: \$11-13k, for 10G vs 25G QSFP devices

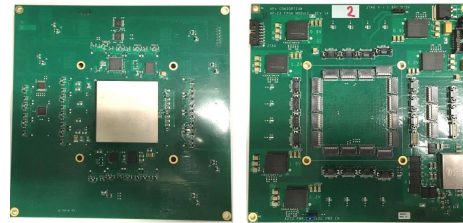
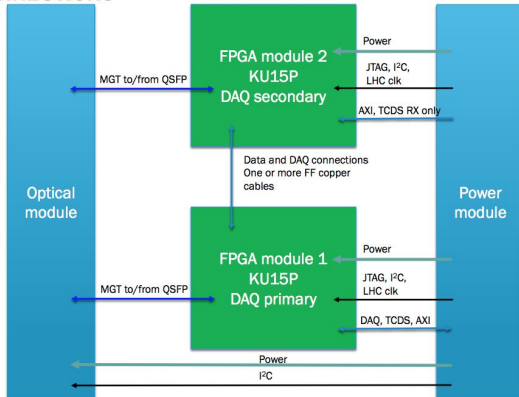
QSFP+ cages
Up to 28 pcs

ZYNQ module

DC-DC converters

Backplane connectors

MODULE CONNECTIONS



X2O KU15P MODULE

Prototypes of all modules produced and tested (power, FPGA, QSFP)
Completely assembled board tested (including thermal tests at full power),
and three units in operation (UF, CERN, UCLA), plus more modules in assembly.

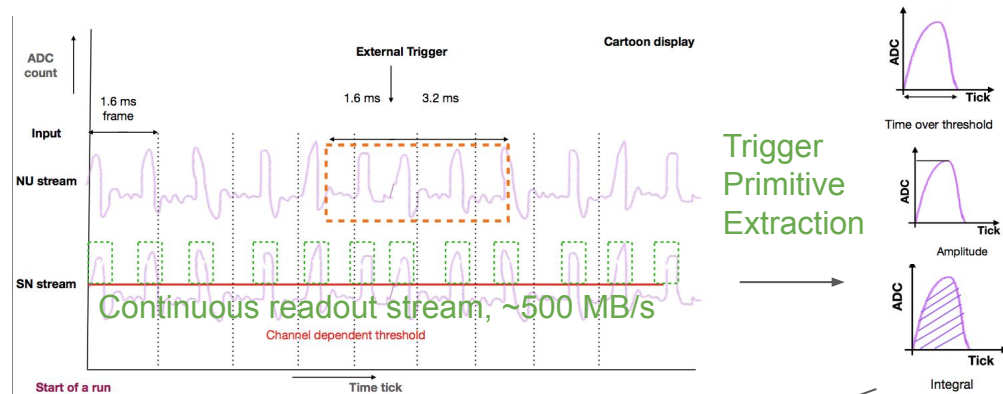
Highlights: Liquid Argon TPC Trigger Development with MicroBooNE & SBND

Planned in situ demonstrations of TPC triggers (including ML based) with currently & soon to be operating MicroBooNE & SBND LArTPCs.

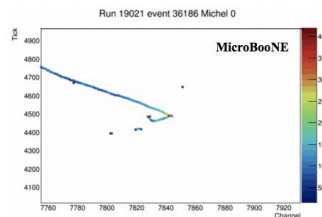
Triggers to be demonstrated first online in CPU/GPU, and then in FPGA over the next 1-2 years.



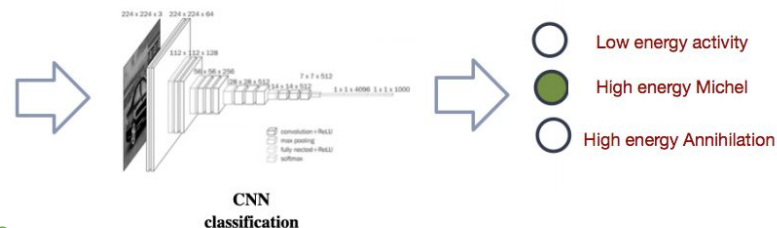
Sample	Train Size	Test Size	Accuracy (%)			Inference Time (ms)
			ϵ_{NB}	ϵ_{LE}	ϵ_{HE}	
NB	12,023	4,027	99.53	0.47	0.12	1.6 \pm 0.1
LE	12,050	3,970	4.01	94.48	1.51	
HE	10,137	3,417	3.63	6.15	90.22	



Higher Level Trigger Decision
(clustering, or image classification)



Trigger Primitive Image



CNN
classification

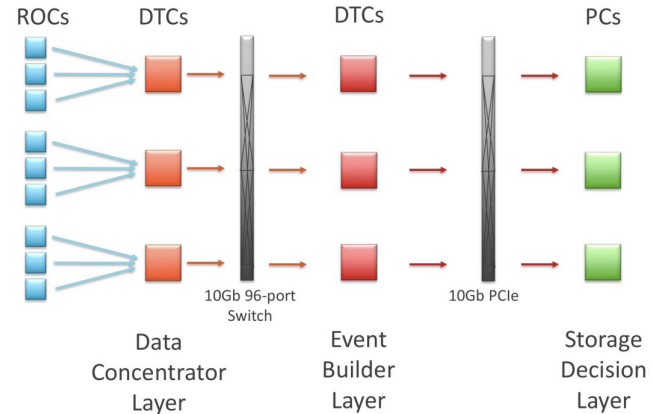
Highlights: Trigger and DAQ for the Mu2e-II experiment

Physics motivation:

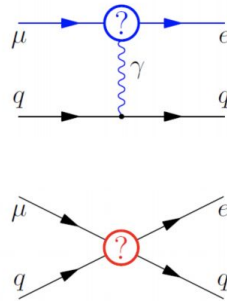
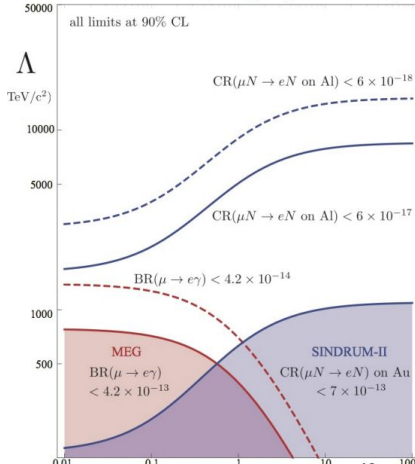
Improve sensitivity to charged-lepton flavor violating (CLFV) neutrino-less conversion of a nuclear bound muon into an electron by an order of magnitude over Mu2e.

- Also 10x radiation levels and 20x data rate!

Mu2e-1 Trigger DAQ consists of multiple layers starting with Readout Controllers



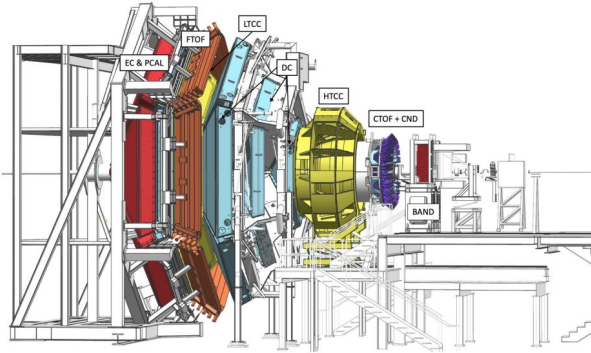
Derived from A. de Gouvea, P. Vogl, Prog. Part. Nucl. Phys. 71 (2013) 75



Mu2e-II Trigger DAQ Options: Trigger-less TDAQ, GPU co-processors, FPGA pre-filtering, FPGA pre-processing and trigger primitives

- Trigger-less TDAQ - requires 10x more hardware
- HLT parallelization non-trivial!
- L1 Hardware trigger implemented on FPGAs utilize HLS
- 2-level TDAQ system running on FPGAs

Highlights: Streaming DAQ system for CLAS12 Forward Tagger



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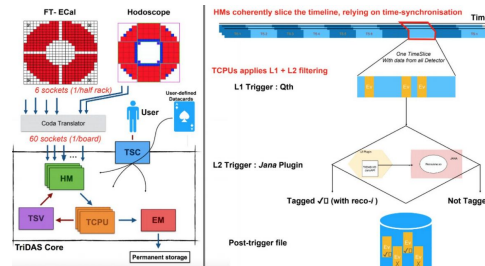
Installed at JLAB Hall-B

- Investigation of the structure of the proton and neutron in ground and excited states

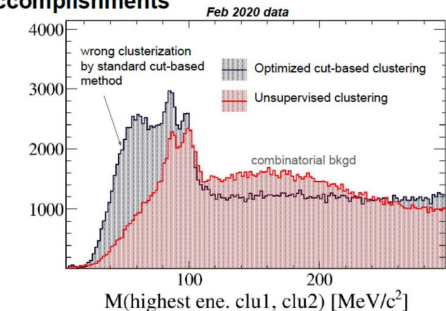
Recent progress towards testing streaming read out of data with full TriDAS chain to study performance

- Must reduce 50 MB/s to ~4 MB/s
- Perform reconstruction in real time
- Full chain tested successfully
- New AI algorithm improves mass resolution and real time clustering

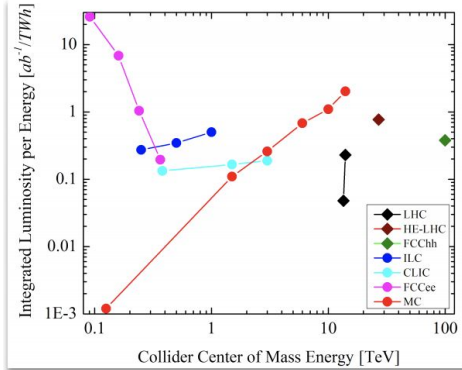
Prototype is being used as the basis for developing a large system for the entire CLAS12 detector



Accomplishments



Highlights: Detector for a Muon Collider (TDAQ)

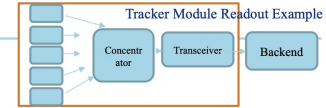


Physics potential for a Muon Collider is formidable, but so is the Beam Induced Background (BIB) Due to the Muon's short lifetime

Data rates and timeline should allow for streaming readout, some R&D needed for FE chips

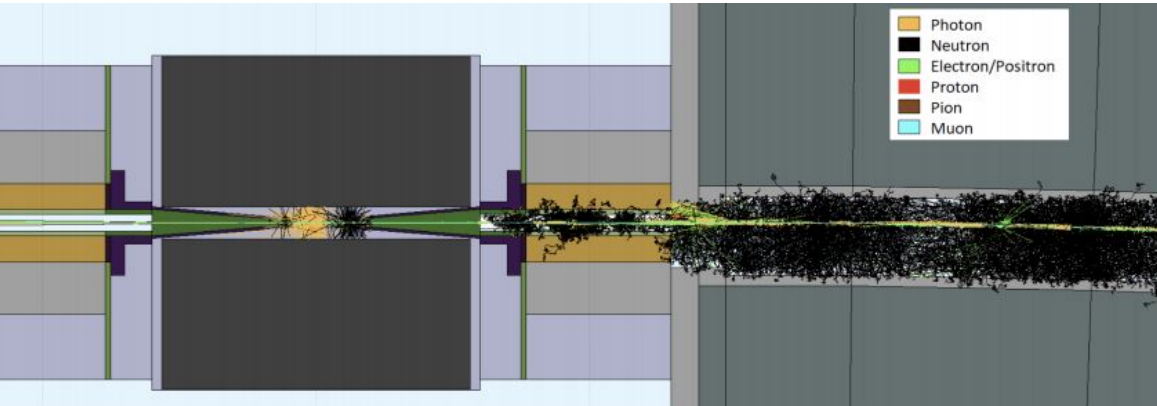
Readout/DAQ Considerations

- Data \Rightarrow bandwidth \Rightarrow power
 - Note time between collisions is 10 μ s = 100 kHz
- Assuming module size of 20 cm^2
 - With 50x50 microns pixel size, get ~800k pixels per module with 1ns window
 - With 1% occupancy, this is up to 8k hits per module in the inner vertex tracker
 - 32 bits to encode x/y/amp/time
- Data rates: 8k hits * 32 bit * 100 kHz * 2(safety factor) ~ 50 Gbps per module (20 cm^2) ~10 Gbps per FE Chip (4 cm^2)
 - Double compared to HL-LHC FE chip. Requires R&D.
- More online handles should be explored: Data compression, some front-end clustering, pT-module based suppression (preliminary estimates indicate x5 rate reduction)
- Downstream electronics needs to be able to accommodate this bandwidth



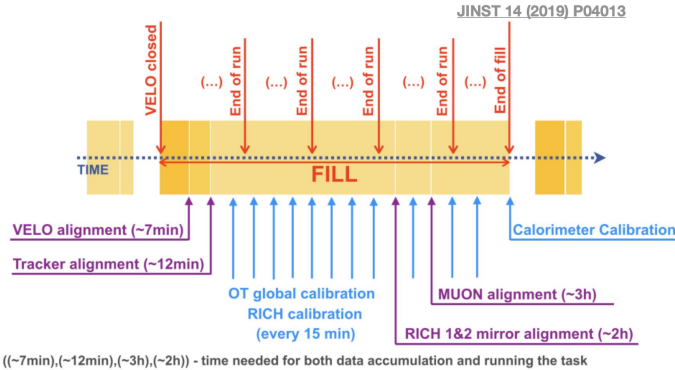
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S. Jindariani, CPAD 2021



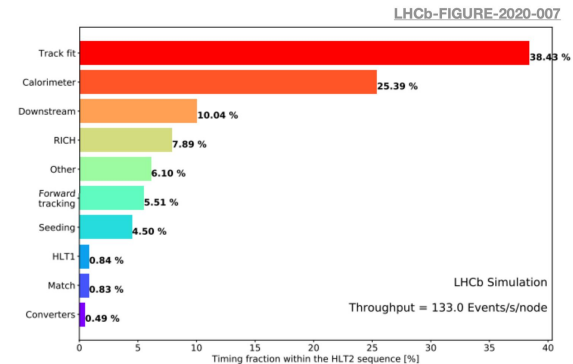
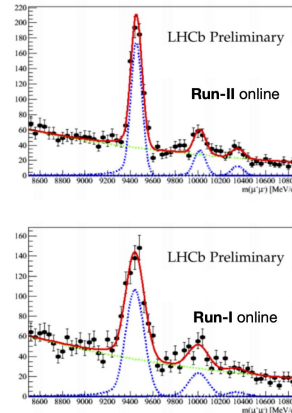
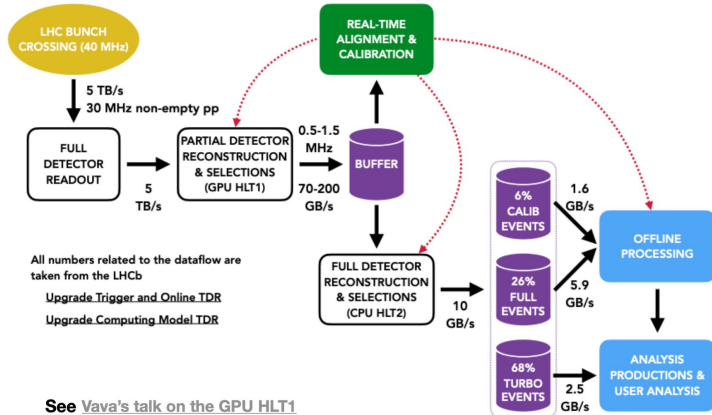
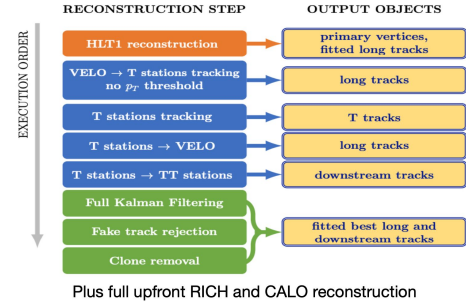
- Data rate for tracker alone are 30 Tbps with 1ns readout window
- Calorimeters will contribute ~the same
- 100-200 readout boards at 10-20 Gbps

Highlights: Real time analysis in Run 3 with the LHCb Experiment



LHCb streaming DAQ system requires continuous calibration online in order to reconstruct offline-quality objects in real-time

- Developing a global calibration system which will run every 15 minutes during the fill



Highlights: Towards an Interpretable Data-driven Trigger system for High- Throughput Physics Facilities (Pie-in-the-Sky talk!)

Envisioning a Self-Driving trigger system

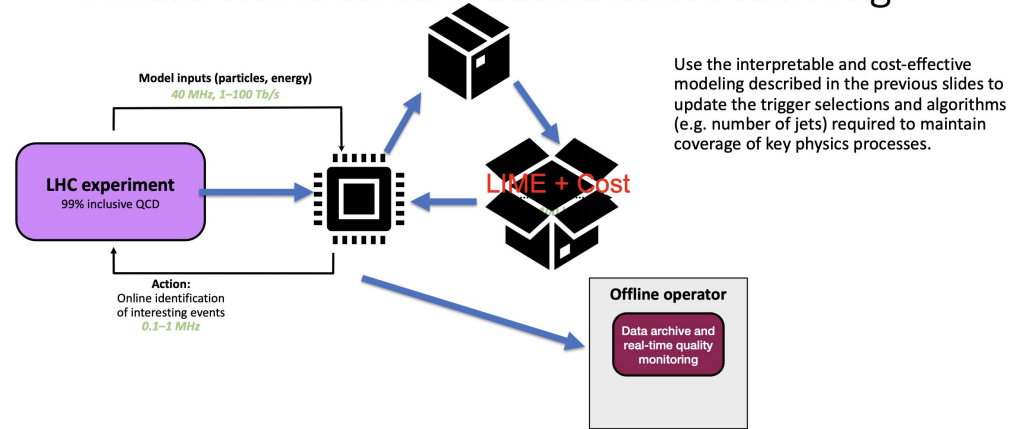
What has been learned such that an update is merited?

- Interpret the output of the algorithm
- “Why” was the event triggered?
- What trigger algorithm was “most important” to the trigger decision?

What are the impacts of those updates?

- Given a definition of the resource cost of a set of triggers, how can we optimize the algorithm execution and usage to minimize that resource usage?
- Cost might include bandwidth considerations, CPU time, data preparation, etc

Future work: stream-based active learning



Goal to minimize the “cost” (Rate!)
... While maximizing the performance!

- Study performed using CMS open-data
- Future work to implement stream-based active learning

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




↓ ATLAS TDAQ Phase-II TDR

 UNIVERSITY OF OREGON

Highlights: The Challenges of Machine Learning at the Edge

Why do we need Real-Time ML ?

New generation sensors and detectors have :

- More pixels; 
- Faster sampling rates; 
- Better sensitivity and dynamic range; 
- Larger surface area; 
- Too much data! 

© Dan Plucinski



Traditional programming

Input

+

Program

=

Output

Machine Learning

Input

+

Output

=

Program

Recognizing patterns

Recognizing anomalies

Non linear regression
→ reconstruction

Faster more flexible programming

Lower computational burden

Fast inference
Low latency decision

Benefits of ML

Highlights: SONIC: coprocessors as a service for accelerated inference of DL algorithms

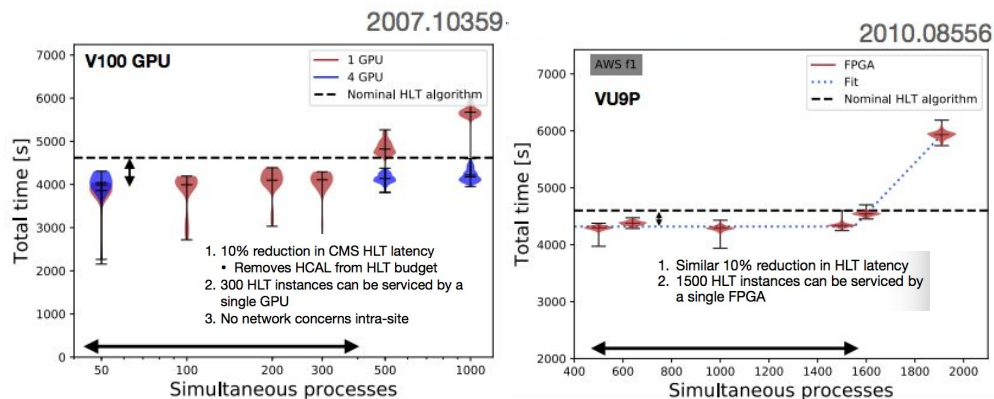
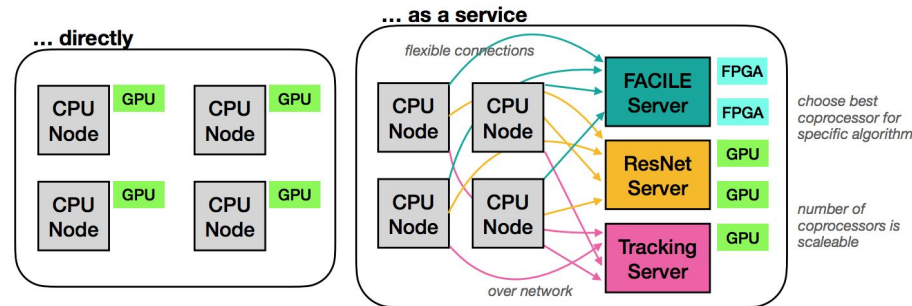
SONIC: Services for Optimized Network Inference on Coprocessors Framework for integrating GPUs and FPGAs as a service (aaS) into physics workflows

Case studies of integrating GPUs/FPGAs aaS into:

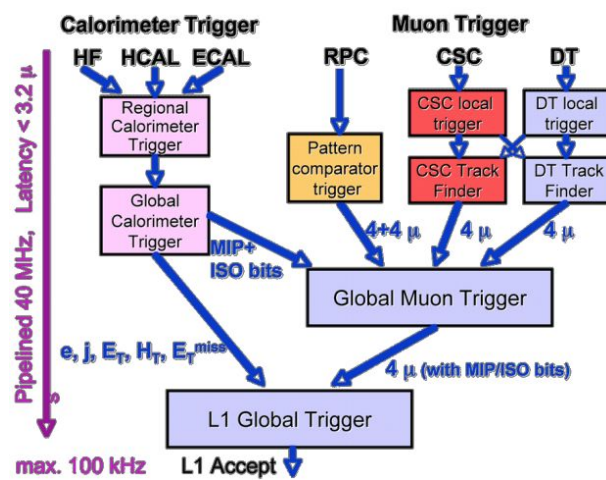
- LHC experiments: GPU, FPGA
- neutrino experiments: ProtoDUNE
- Gravitational waves: LIGO denoising

As-a-service paradigm introduces coprocessors to HEP with minimal changes to pre-existing computing workflows

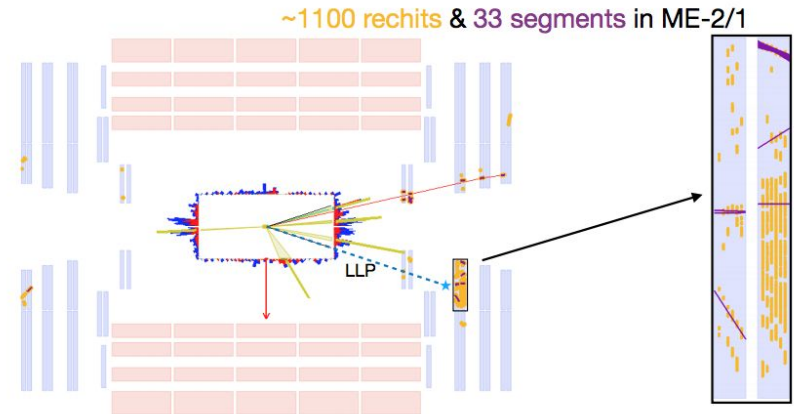
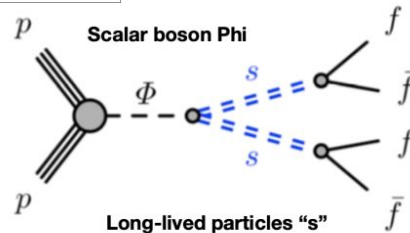
Demonstration of scaled CMS HLT speed-up with hadron calorimeter reconstruction (low- latency, high batch) performed on GPUs and FPGAs



Highlights: Triggering on Long-Lived Particles decaying to Hadronic Showers in CMS Muon System



Development of new trigger path targeting long-lived particles in CMS, using the muon system as a sampling calorimeter.
 Dedicated trigger for showers in CSC could extend CMS's reach for LLPs to $>1\text{m}$ lifetime.
 Can improve acceptance by 20-30x; on-track to be brought online in CMS Run 3.



Highlights: hls4ml enabling real-time deep learning in particle physics

- **hls4ml** is a library for automatic translation of deep learning models to FPGA firmware for inference with ultra low latency

- **First target applications:**
hardware trigger of LHC experiments and detector front-end electronics

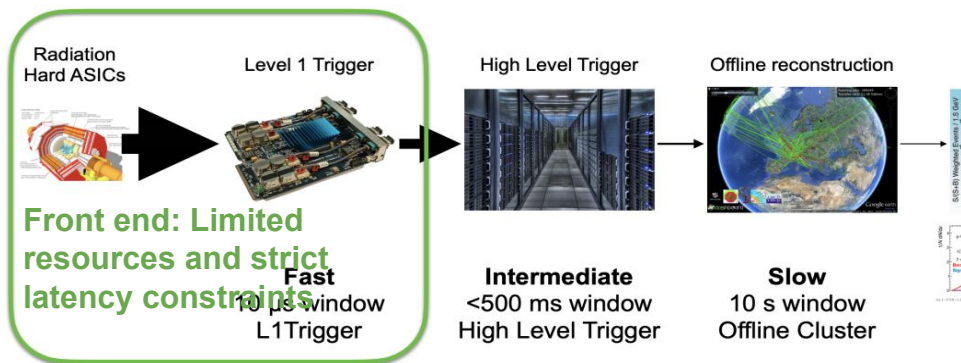
Recent developments/library expansions:

Quantization-aware training and pruning [[arXiv:2006.10159](https://arxiv.org/abs/2006.10159)]

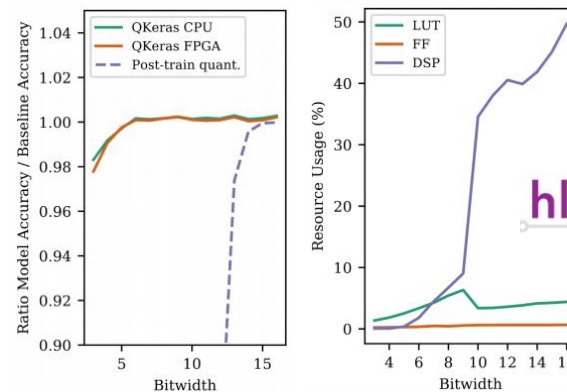
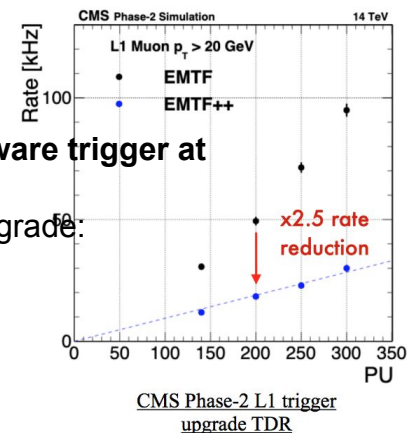
Convolutional neural networks [[arXiv:2101.05108](https://arxiv.org/abs/2101.05108)]

Custom architectures as graph neural networks [[arXiv:2008.03601](https://arxiv.org/abs/2008.03601), [arXiv:2012.01563](https://arxiv.org/abs/2012.01563)]

+ more...



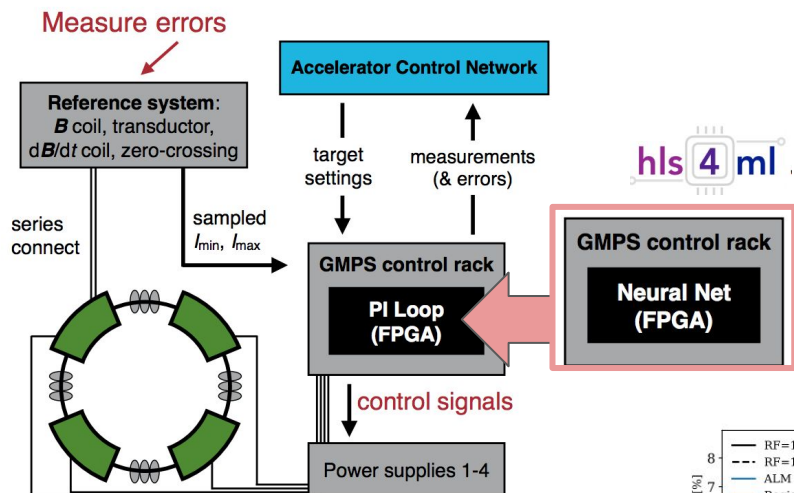
Applications for the hardware trigger at LHC experiments, e.g. CMS Phase-2 L1 trigger upgrade:



Highlights: Real-time AI for Accelerator Control

A study at the Fermilab Booster:

Megawatt proton beam with high proton per pulse density and minimal beam losses is required to meet DUNE requirements → use of ML regulator to enhance beam control [[arXiv:2011.07371](https://arxiv.org/abs/2011.07371)]

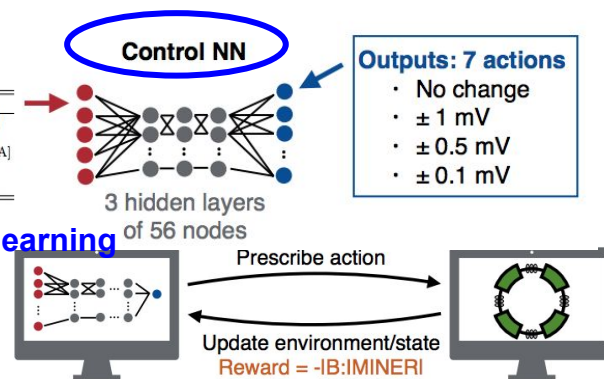


Can naturally incorporate many inputs.
Offers potential for "live" adjustments to the algorithm parameters while in operation.

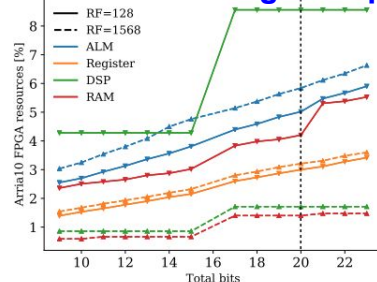
Inputs: current values for the five important signals

Parameter	Details [Units]
B:IMINER	Setting-error discrepancy at injection [A]
B:LINFRQ	60 Hz line frequency deviation [mHz]
B:VIMIN	Compensated minimum GMPS current [A]
I:IB	MI lower bend current [A]
I:MDAT40	MDAT measured MI current [A]

+ reinforcement learning



+ design footprint minimization



Comfortably fit within 6% of the target Arria10 FPGA's resources.

Can trade serial / parallel designs to trade resources for latency.

reuse factor	DSP	BRAM	MLAB	ALM	Register	Latency
128	130	114	229	21.4k	51.2k	2.8 μs
224	74	100	1420	40.2k	78.3k	4.1 μs
1568	26	38	357	24.9k	54.9k	17.2 μs
Available	1518	2713	...	427k	1.7M	...

Appears to perform better on historical booster data

Deployment planned for this spring for in situ demonstration

Final Thoughts

Mastering extreme environments and data rates in HEP experiments:

Future HEP detectors will involve extreme environments and exponential increases in data rates to explore elusive phenomena. ... To do so requires the intimate integration of intelligent computing with sensor technology.

[BRN Study Group, Identified Grand Challenges]

Many ongoing efforts reflect **challenge** and need for integrating intelligent computing into TDAQ systems.

BRN TDAQ PRD's:

- 21: Achieve on-detector, real-time, continuous data processing and transmission to reach the exascale
- 22: Develop technologies for autonomous detector systems
- 23: Develop timing distribution with picosecond synchronization