



Workshop da Rede Nacional de Física de Altas Energias (RENAFAE) 2022

Trigger de Múons de Primeiro Nível Assistido pela Calorimetria (TileCal) no Experimento ATLAS

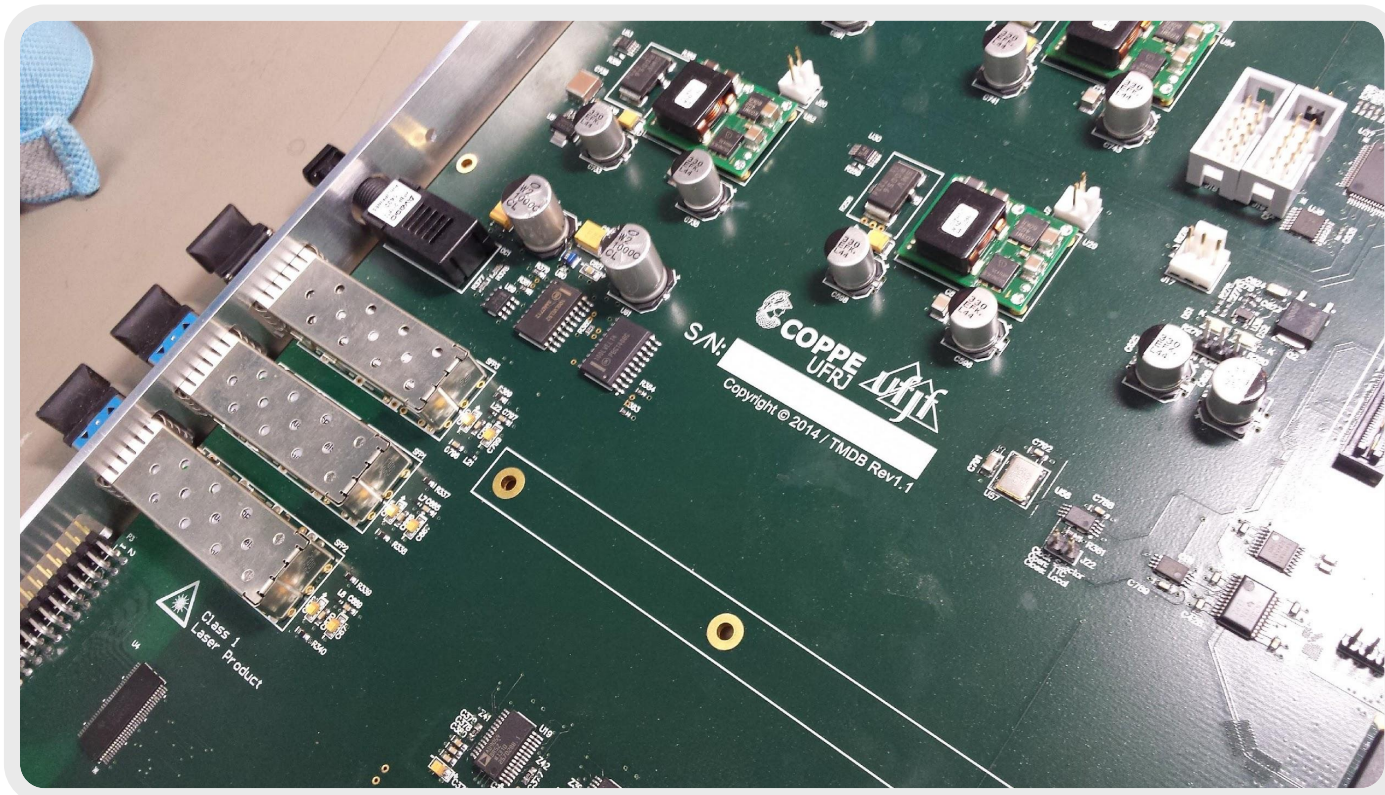
VICTOR A. FERRAZ
on behalf of TMDB team





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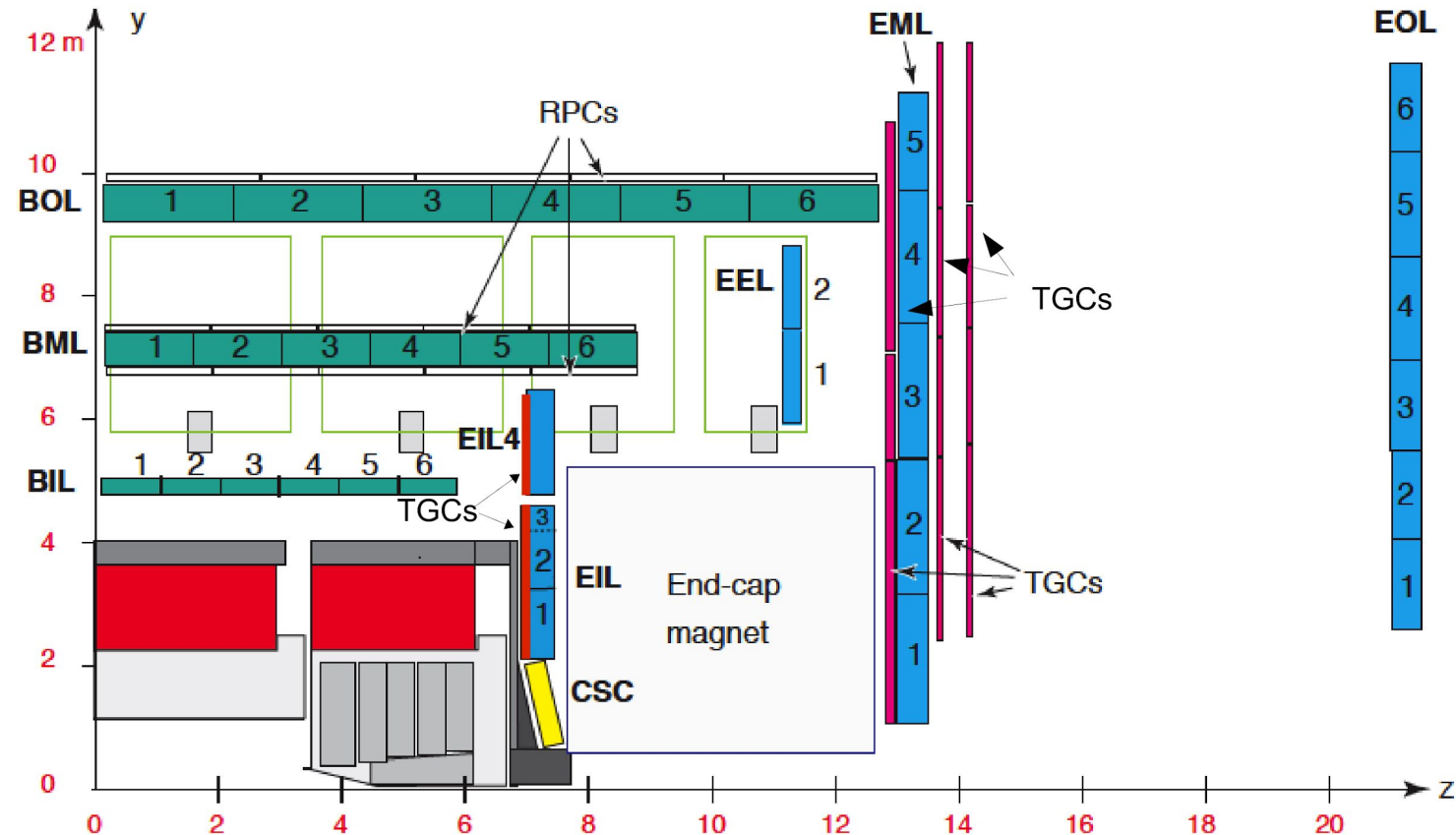




Muon Trigger Assisted from TileCal

Introduction

- ❖ A significant part of the muon trigger rate in the end-caps is background

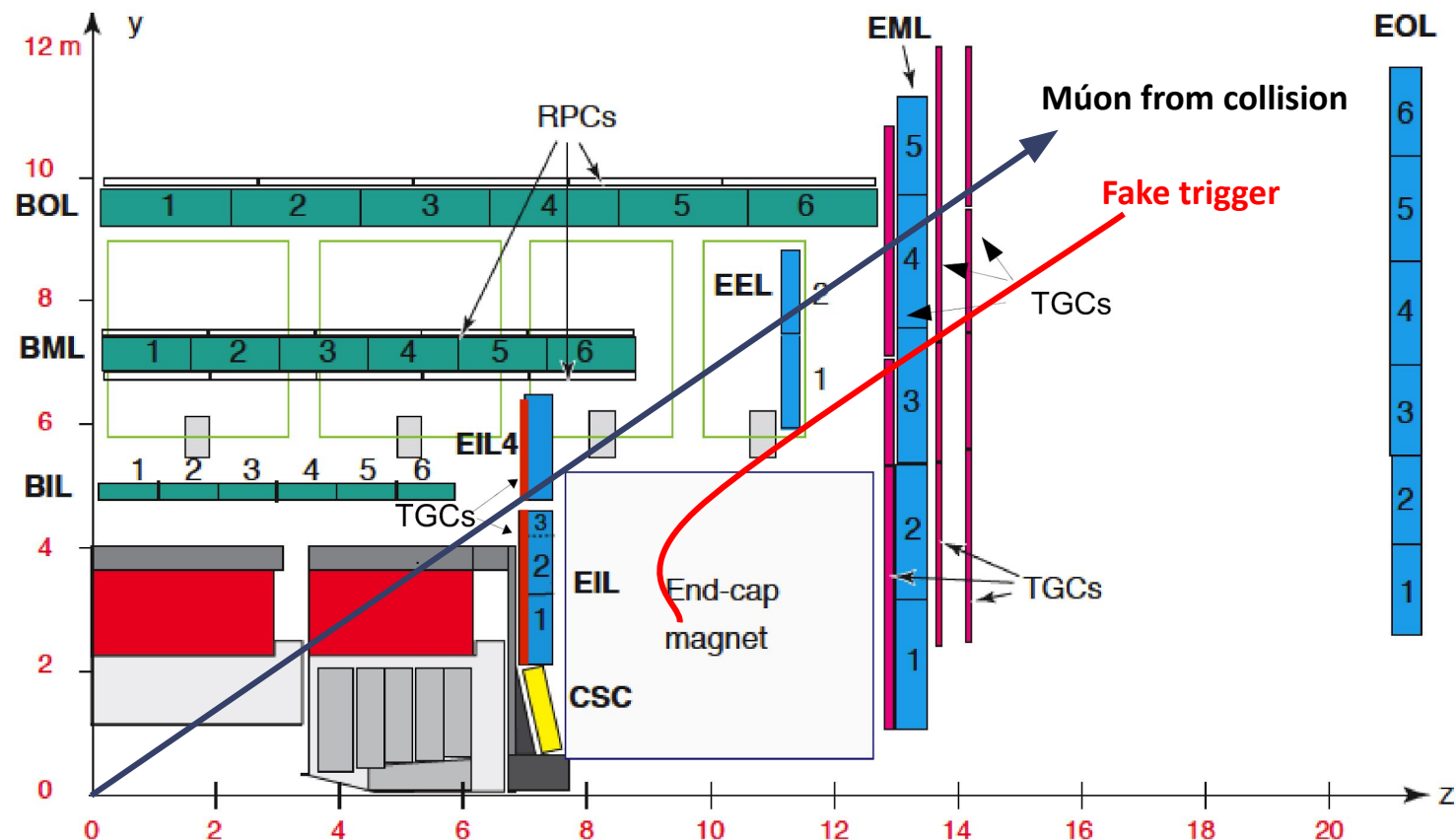




Muon Trigger Assisted from TileCal

Introduction

- ❖ A significant part of the muon trigger rate in the end-caps is background
 - Low energy particles (mainly protons) produce fake triggers by hitting the end-cap trigger chambers





Muon Trigger Assisted from TileCal

Introduction

- ❖ Expected single-muon Level-1 rate (based on 2012 data and 8 TeV) with bunch spacing 25 ns and instantaneous luminosity $3 \times 10^{34} \text{ cm}^{-2} \text{ s}^{-1}$ for online $p_T > 20 \text{ GeV}$
 - From ATLAS TDAQ System: Phase-I Upgrade TDR

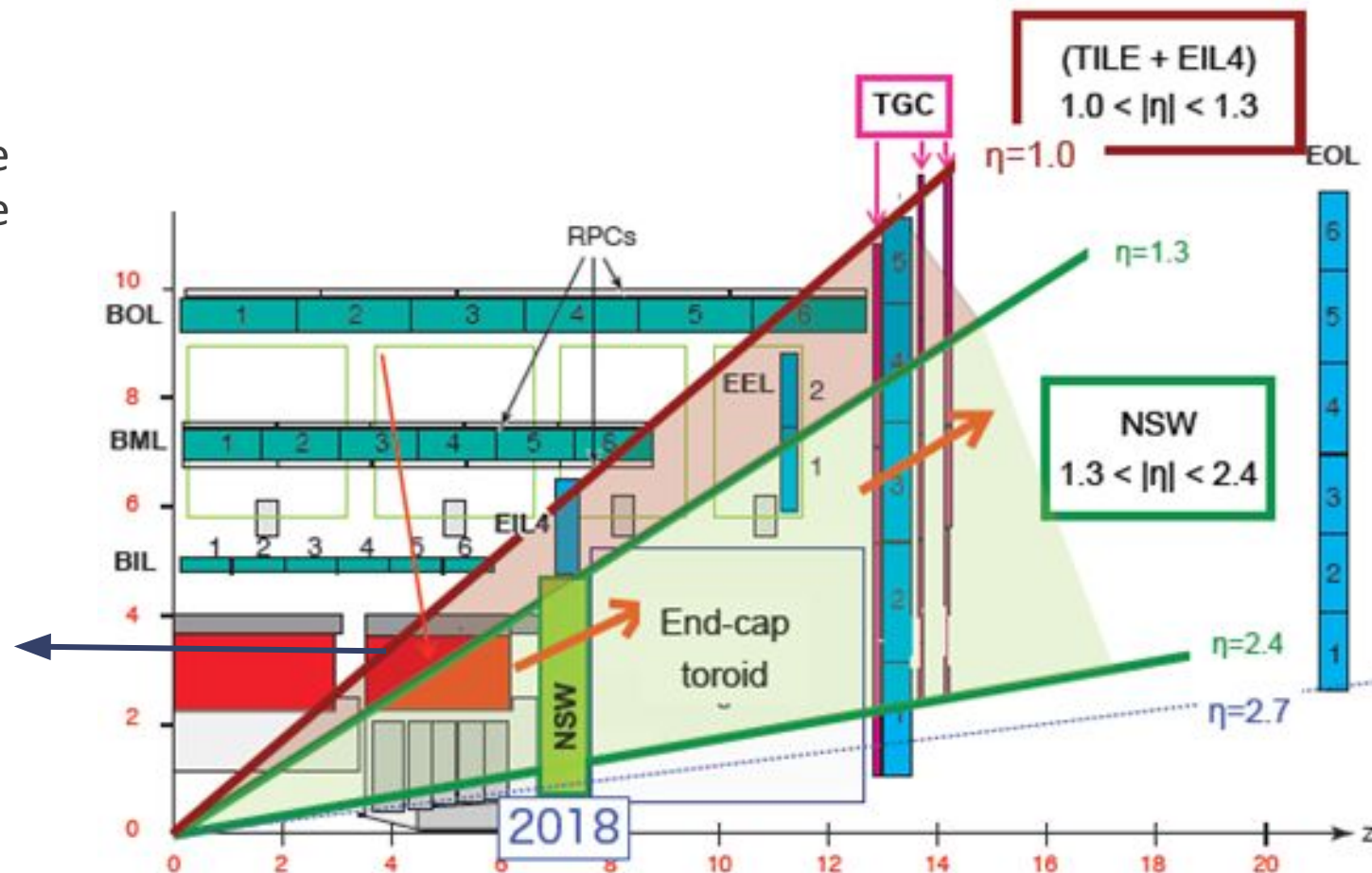
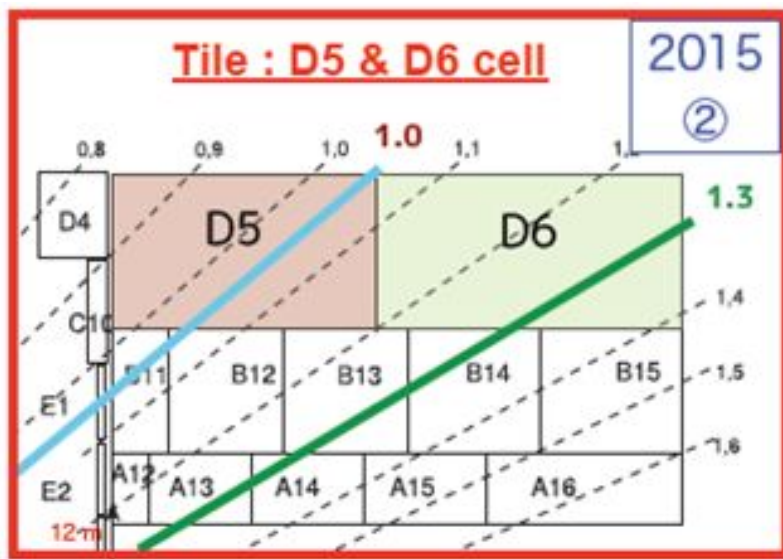
Online $p_T > 20 \text{ GeV}$ $3 \times 10^{34} \text{ cm}^{-2} \text{ s}^{-1}$	No Change	TGC EIL4 + (TGC FI or NSW)	TGC EIL4 + (TGC FI or NSW) + Tile Cal.	TGC EIL4 + (TGC FI or NSW) + Tile Cal. + low field mask
	Rate [kHz]	Rate [kHz]	Rate [kHz]	Rate [kHz]
Run 2 (pre NSW)	51	34	31	28
Run 3 (post NSW)		17	15	13



Muon Trigger Assisted from TileCal

Introduction

- ❖ The **Tile- μ trigger** aim to improve the rejection of fake triggers, by using the energy loss in the TileCal outermost layer (D-layer)

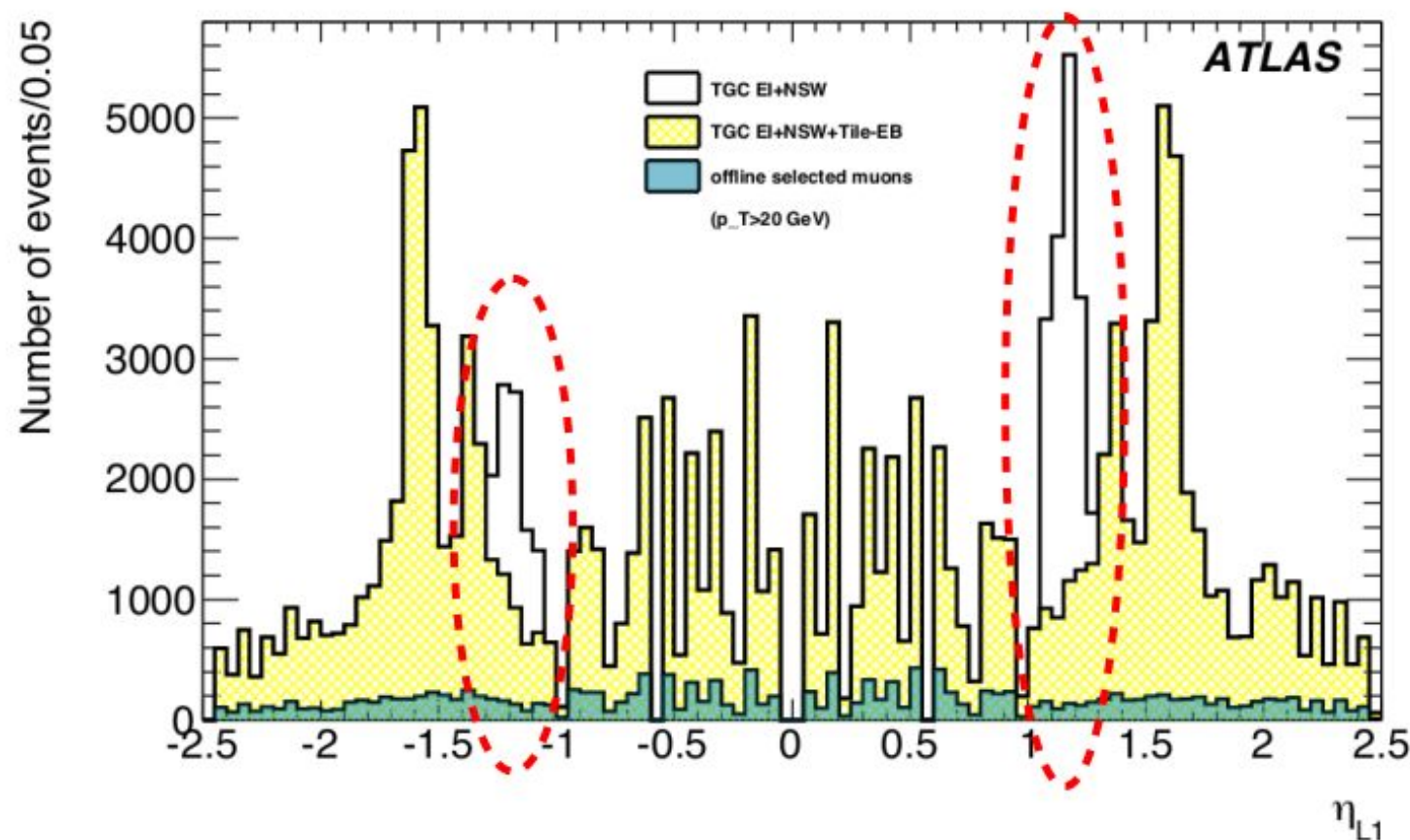




Muon Trigger Assisted from TileCal

Introduction

- ❖ Use Tile Calorimeter D-layers in the region $1 < |\eta| < 1.3$ in coincidence with the TGC inner station chambers
- ❖ By simulation, using a threshold of 500 MeV:
 - 97 % efficiency
 - 17 % rate reduction

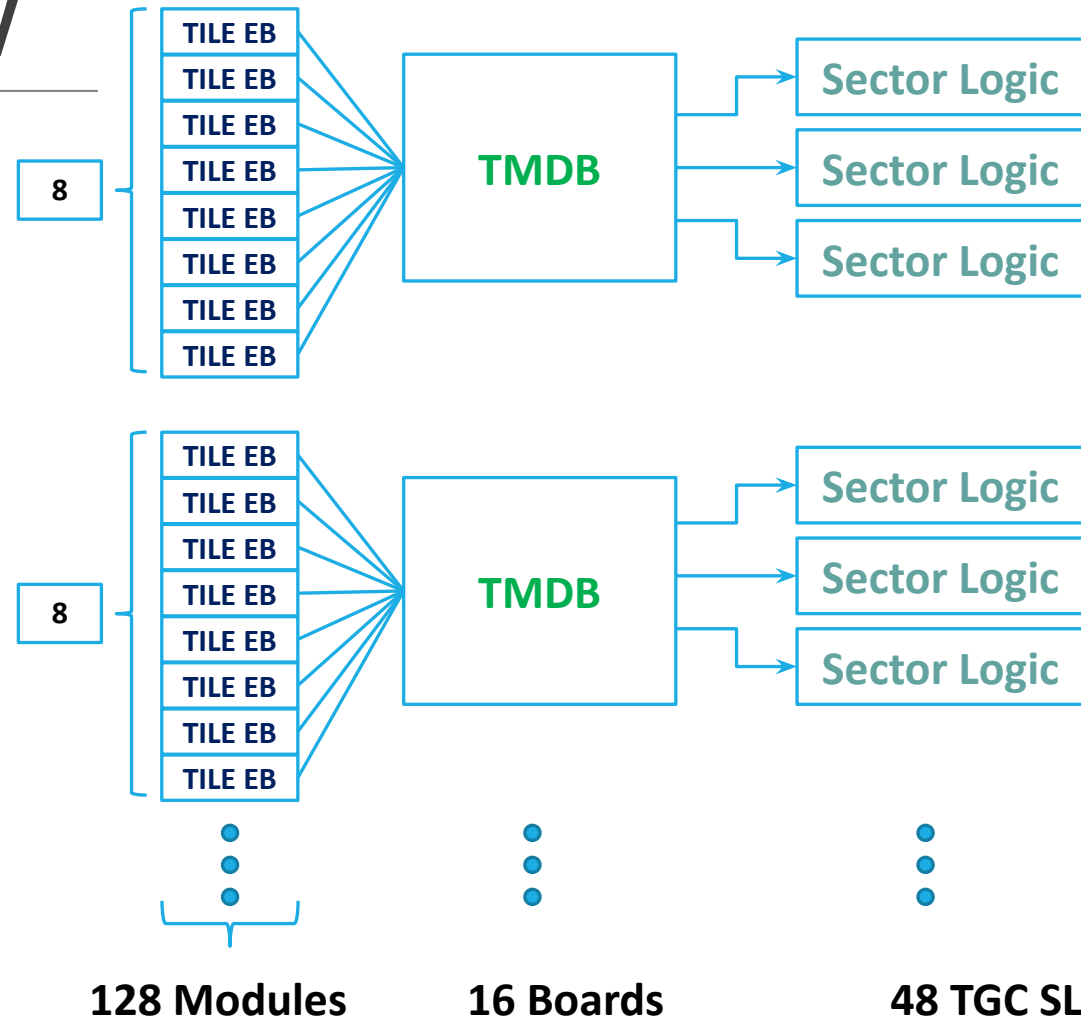




System Overview

TileMuon Trigger

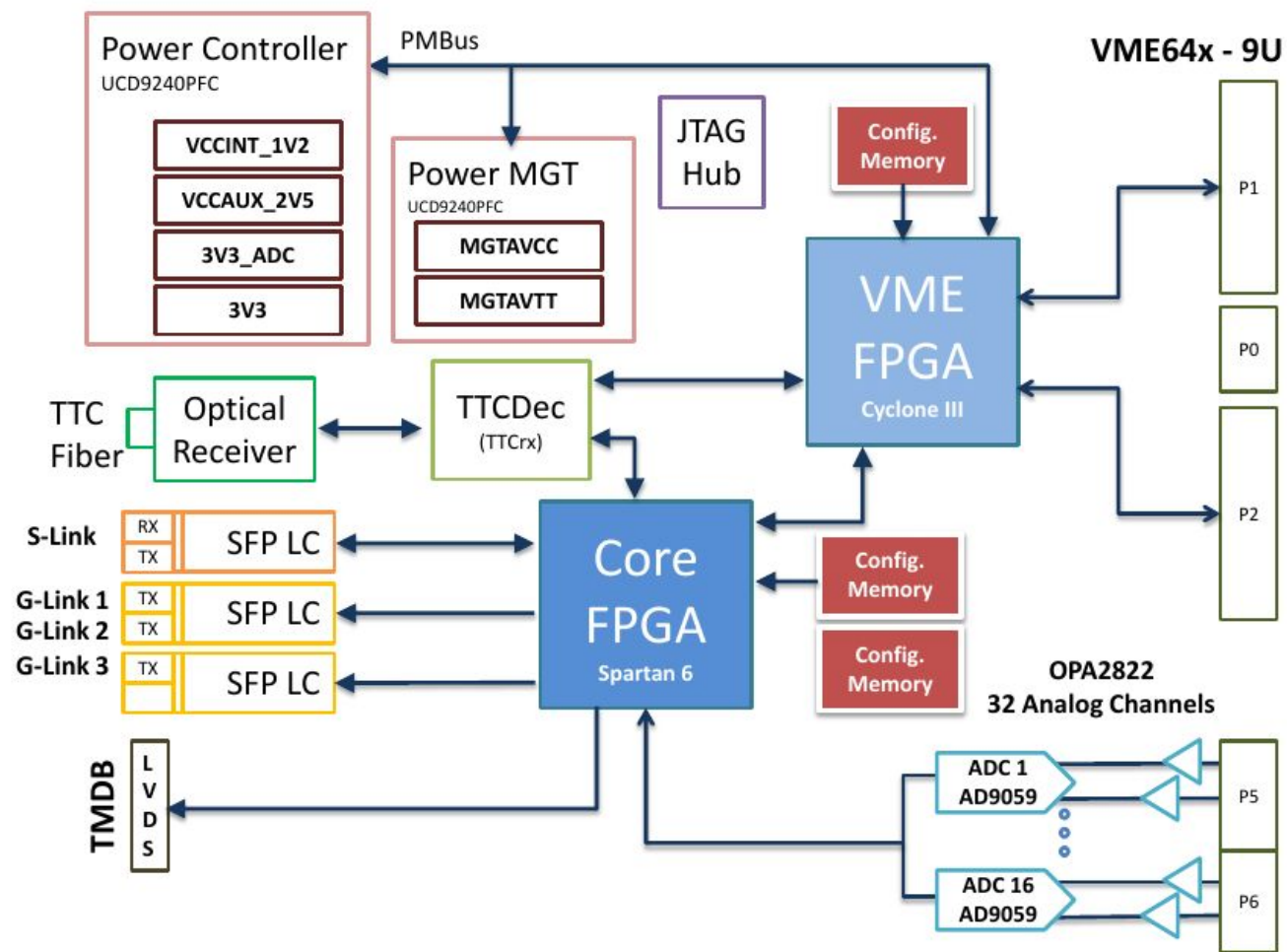
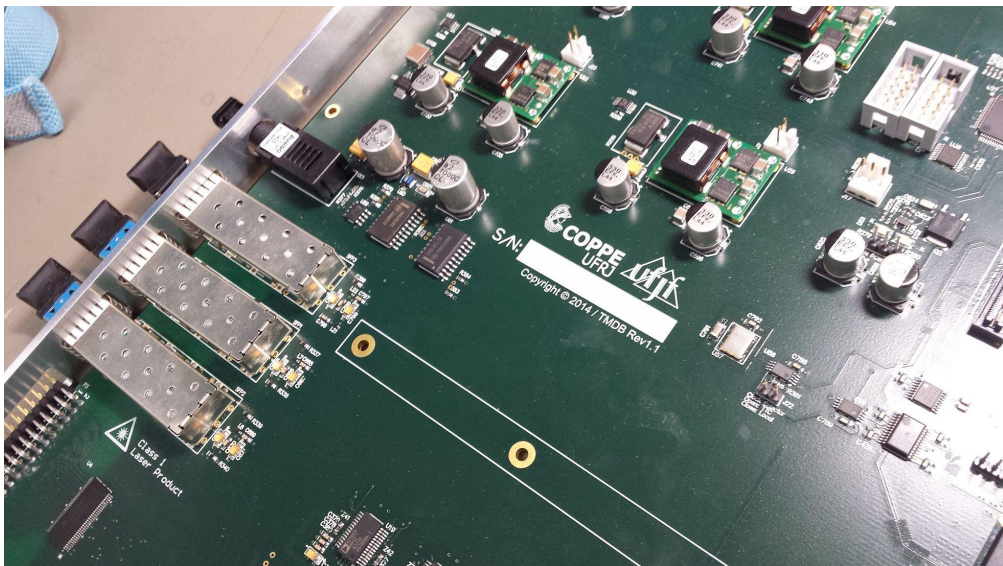
- ❖ 128 Extended Barrel Tile modules
 - 4 channels/module
- ❖ 16 TMDB's
 - 8 Tile modules per TMDB
- ❖ 48 TGC Sector-Logic boards
 - Each TMDB communicates with 3 Sector-Logic boards





System Overview

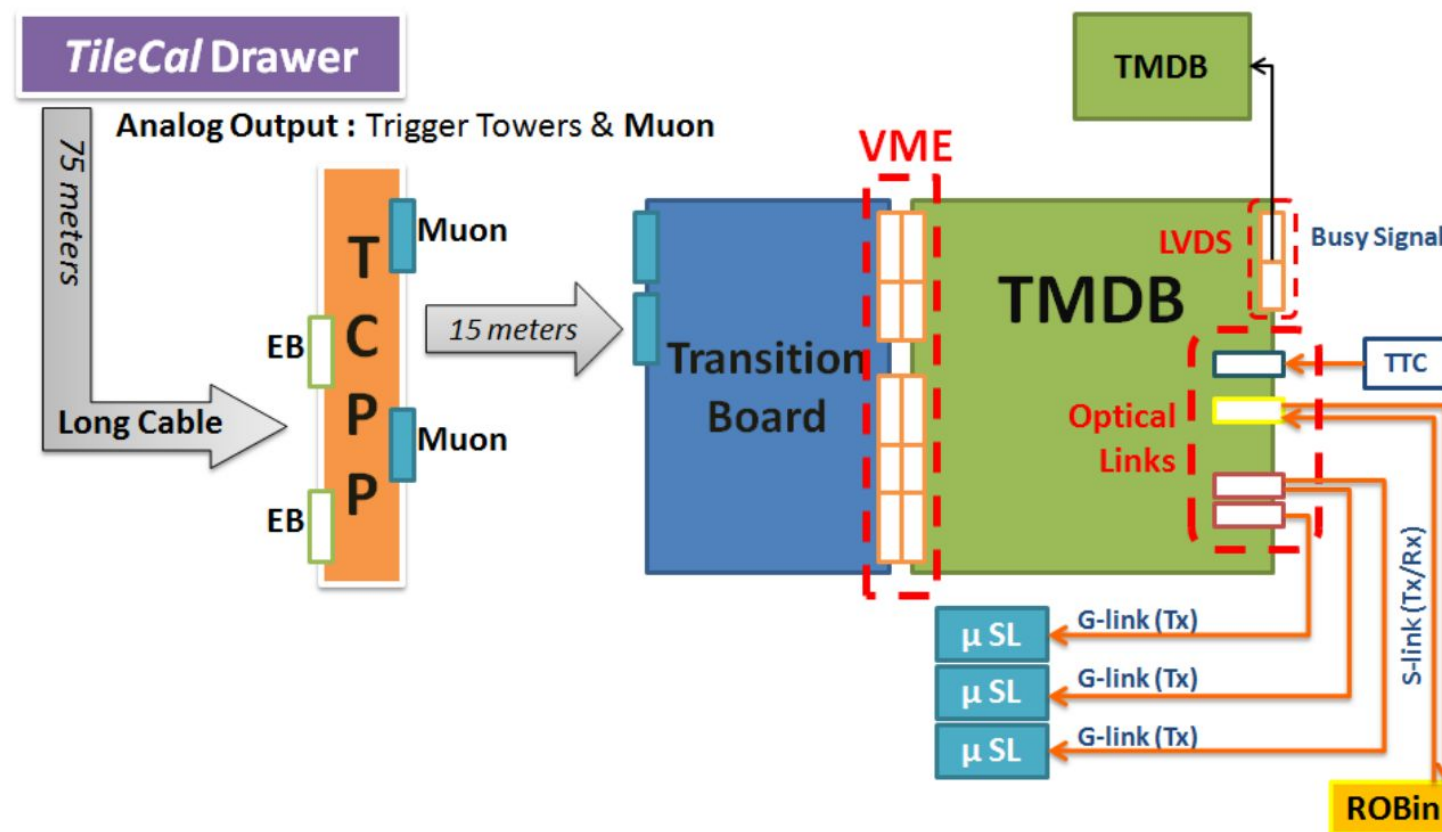
- ❖ Hardware
- ❖ TileMuon Digitizer Board



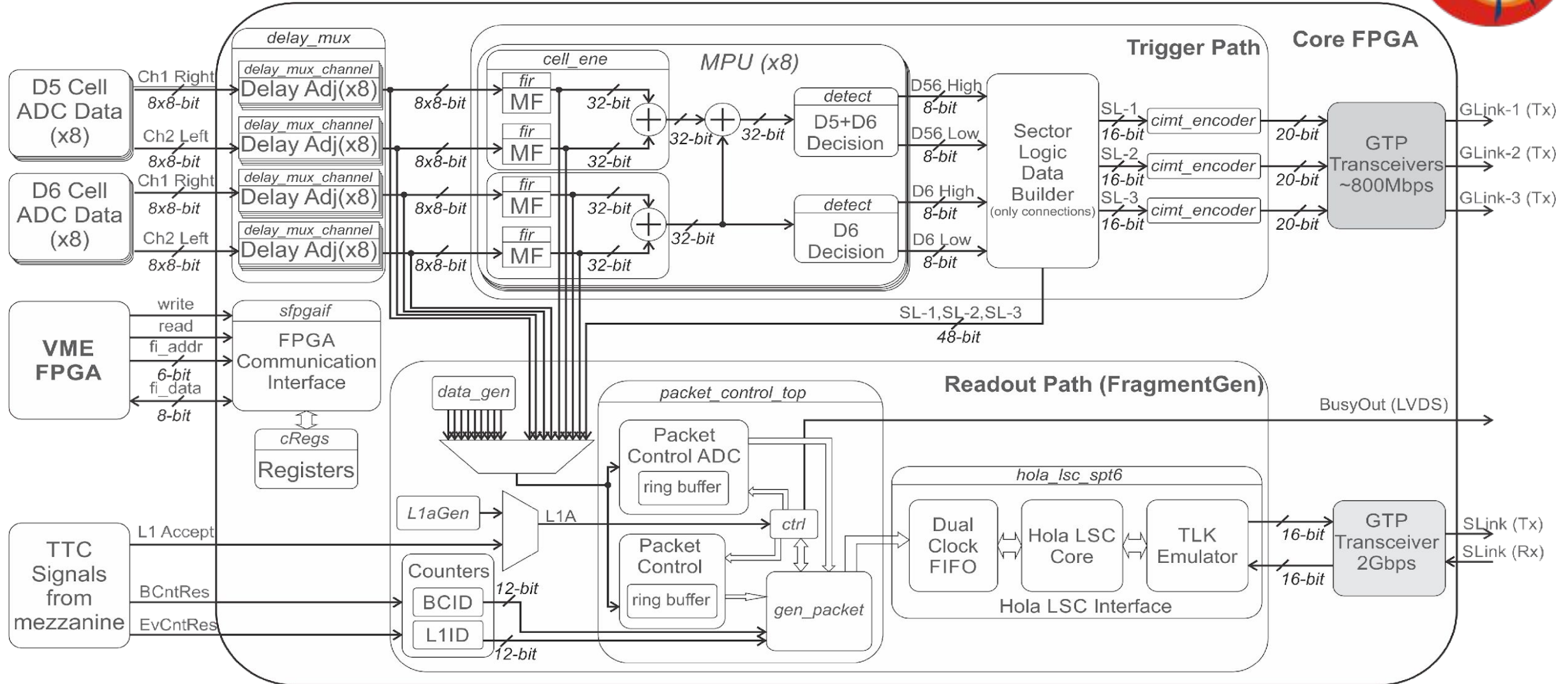


System Overview

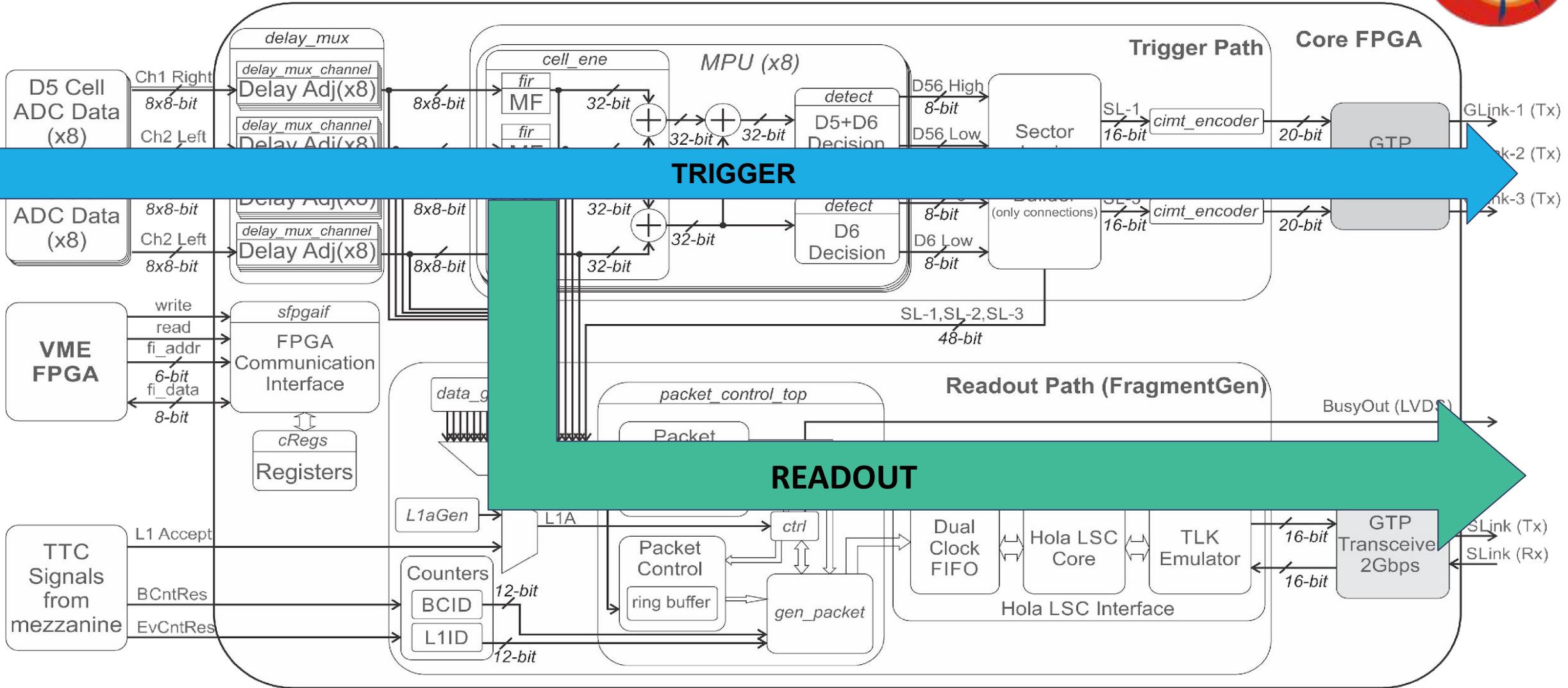
❖ Signal Path



FPGA Implementation: core

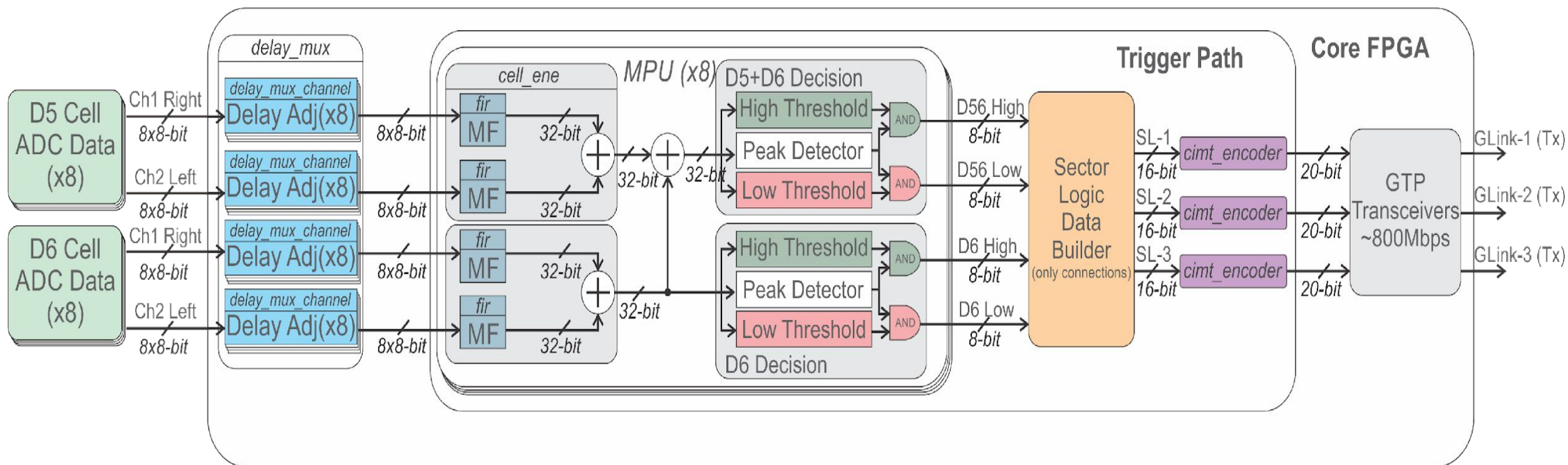


FPGA Implementation: core



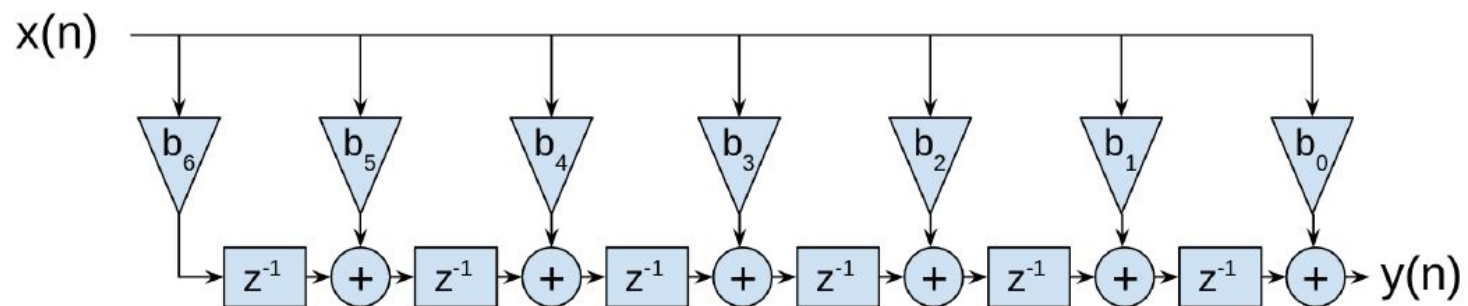
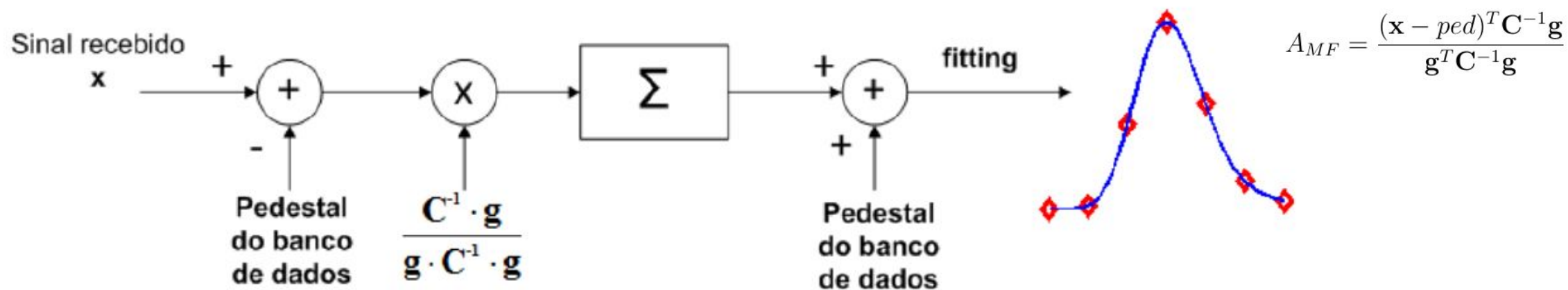


FPGA Implementation: Trigger





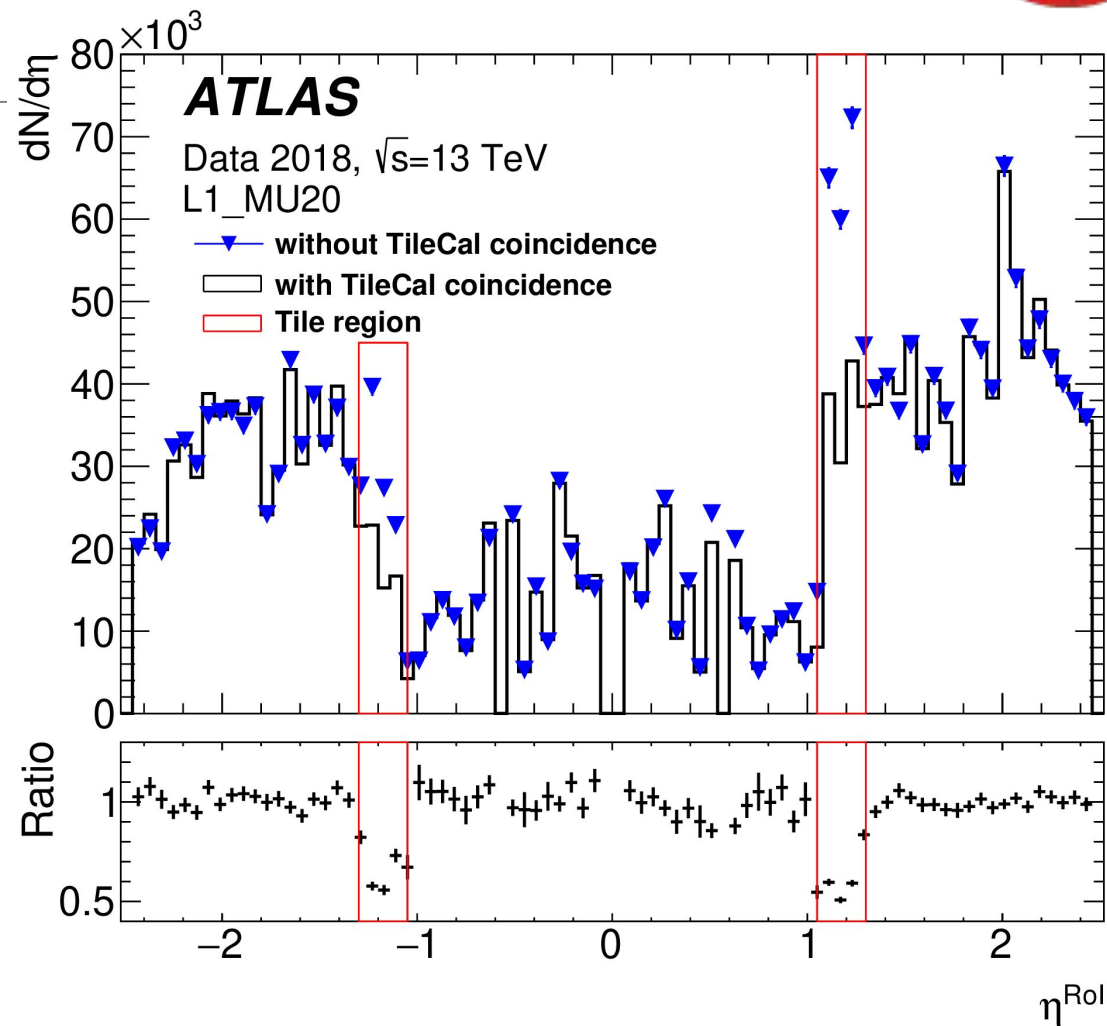
FPGA Implementation: Matched Filter





Operation Results

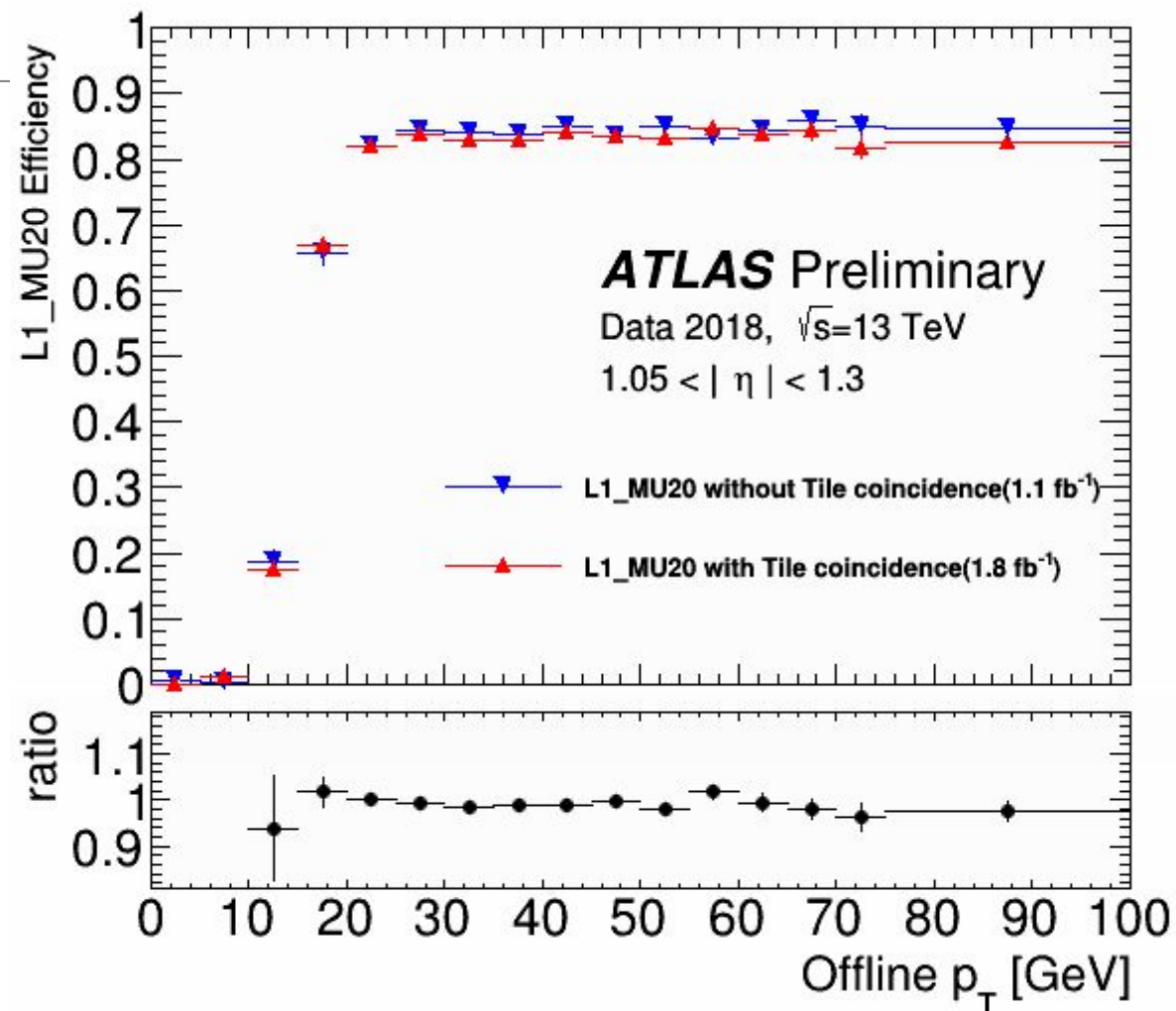
- ◆ The L1 MU20 candidates
- ◆ A comparison between η^{RoI} of all the L1 MU20 entries (white histogram) and the subset which passes the TileCal coincidence requirement (blue histogram) shows the amount of the additional reduction at $1.05 < |\eta^{\text{RoI}}| < 1.3$ by the TileCal coincidence.





Operation Results

- ❖ The L1 MU20 efficiency
- ❖ The trigger efficiency is compared between runs when TileCal coincidence is used in the Level-1 trigger decision (red) and is not used in the decision (blue)
- ❖ Additional inefficiency due to the TileCal coincidence is limited up to: $\sim 2.5\%$

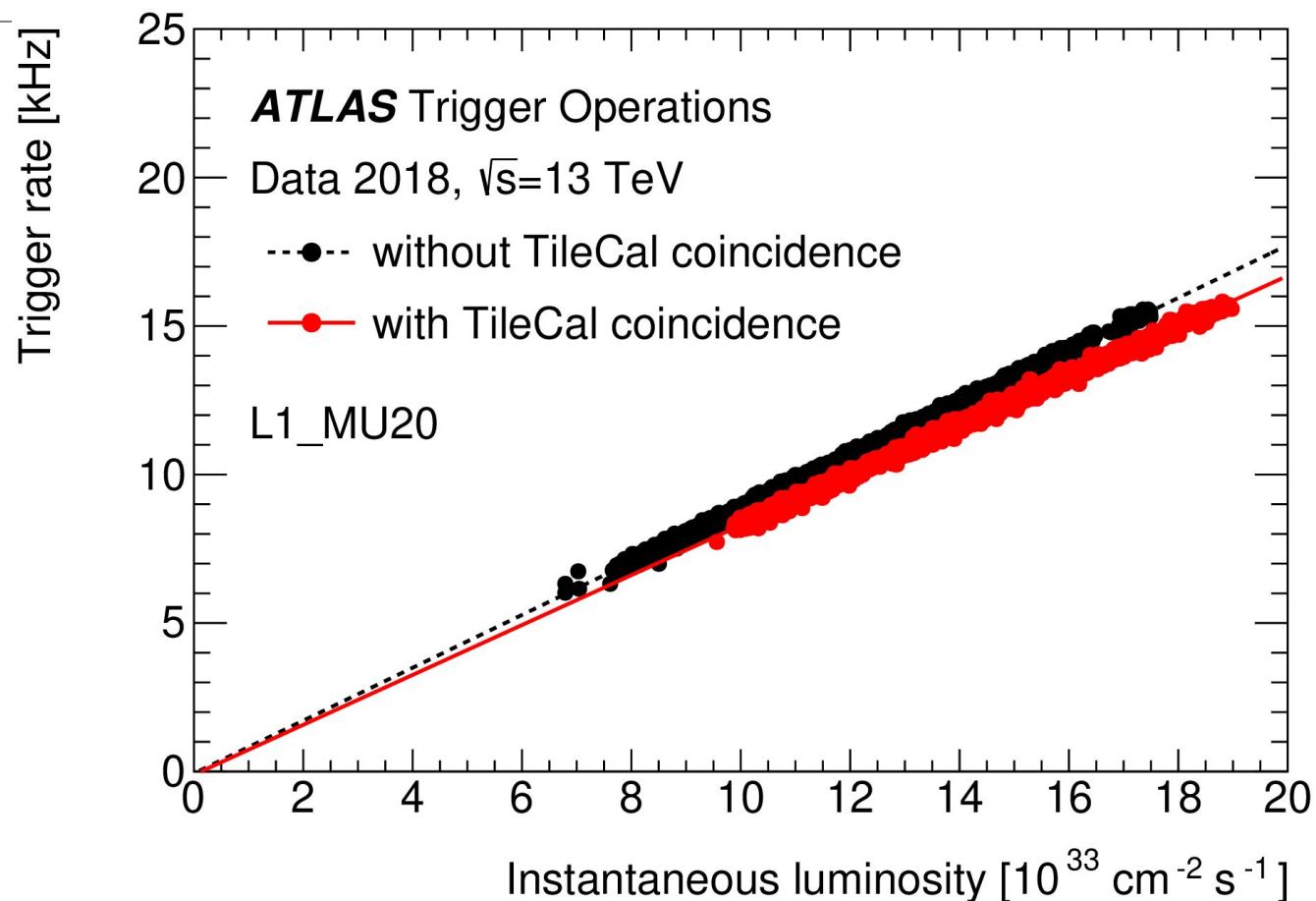




Operation Results

◆ The L1 MU20 trigger rate

- ◆ Coincidence at $1.05 < |\eta| < 1.3$ discards charged particles that only traverse the big wheel, which are originated from secondary interactions in the ATLAS Endcap toroid of the beam backgrounds
- ◆ The comparison of the fit results shows that the reduction of the L1_MU20 trigger rate for all the muon spectrometer coverage is about 6% owing to the new TileCal coincidence requirement





Updates for Run 3

- ❖ New “Firmware”
 - Add Idle words for Sector Logic-TGC communication
 - New Data Fragment for S-Link
 - Inclusion of new information from cell D5 (new MF as well)
 - New Data Frame for Sector Logic-TGC
- ❖ Analysis
 - Testing the wiener filter efficiency
 - Performing Matched Filters calibration (new coefficients, including D5 cells)
- ❖ Extended Applying deep learning/ neural networks to classifying muons
 - Using the offline information to acquire muon event
 - Separating the training/test dataset
 - Implementing deep learning (CNN and RNN - LSTM (Long Short-Term Memory)
 - Increasing statistics