

Development status of Ultra-low latency online event selection trigger system in COMET Phase - I

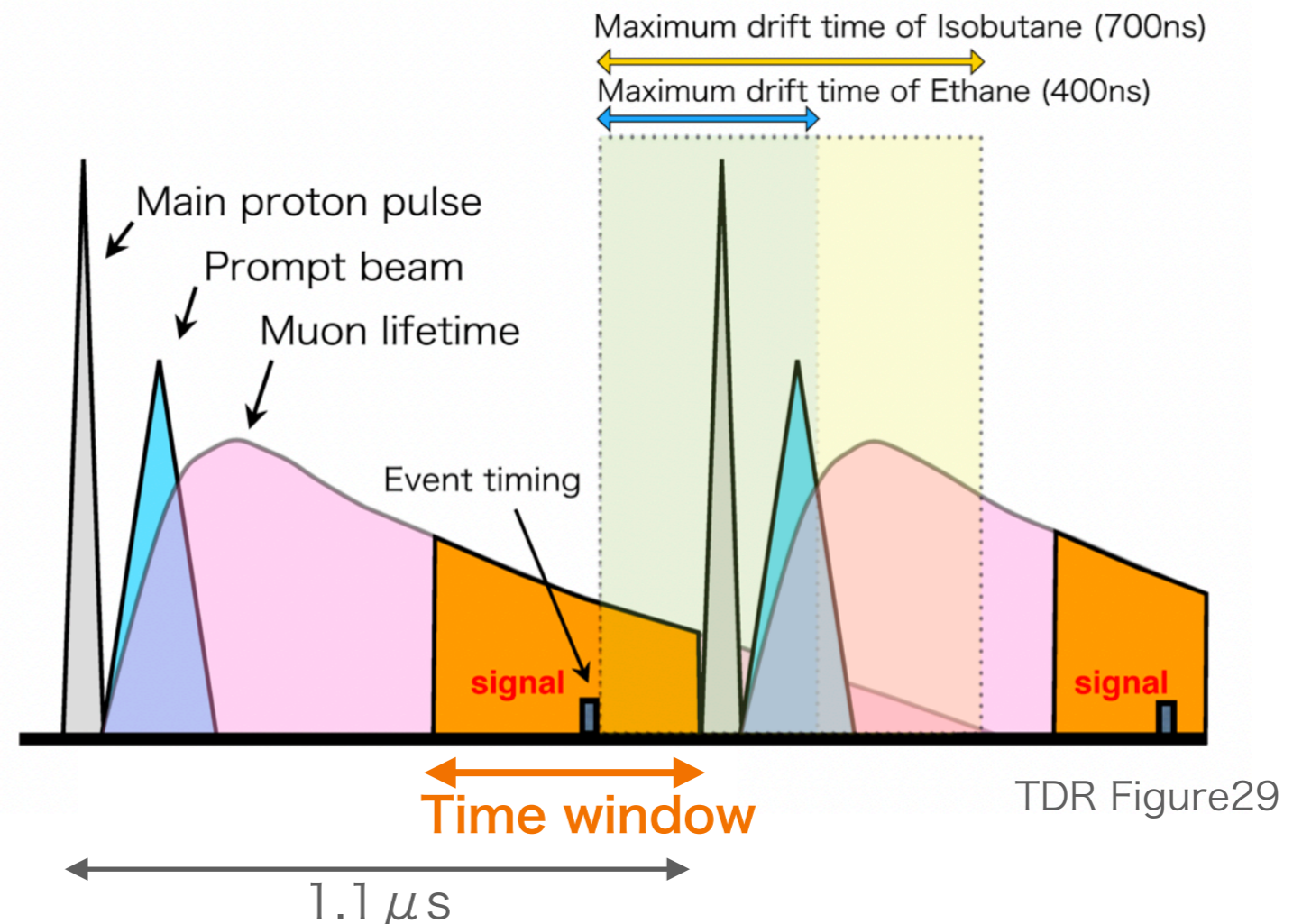
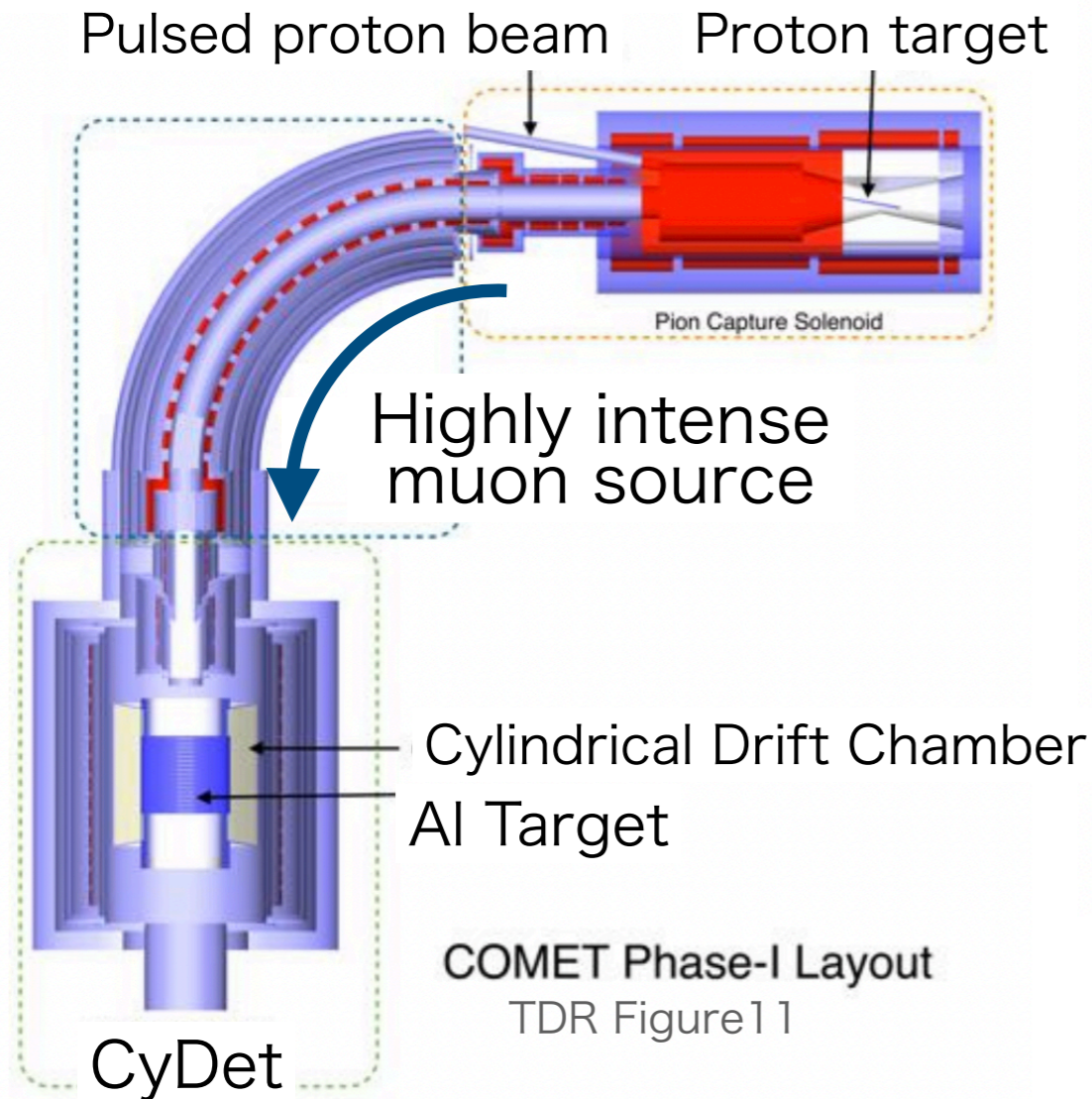
Masaki Miyataki

2022/12/22

Contents

- COMET Phase-I
- Online trigger system
- CDC trigger chain test
- CyDet trigger chain test
- New online event classification algorithm

COMET Phase-I experiment



- **Purpose** : Search for μ -e conversion in an Al target
 - Signal : monoenergetic 105 MeV electron
- **Single event sensitivity** : 3.0×10^{-15} (100 times the current sensitivity)
- **Detector** : Cylindrical detector system
 - electron momentum and timing measurements

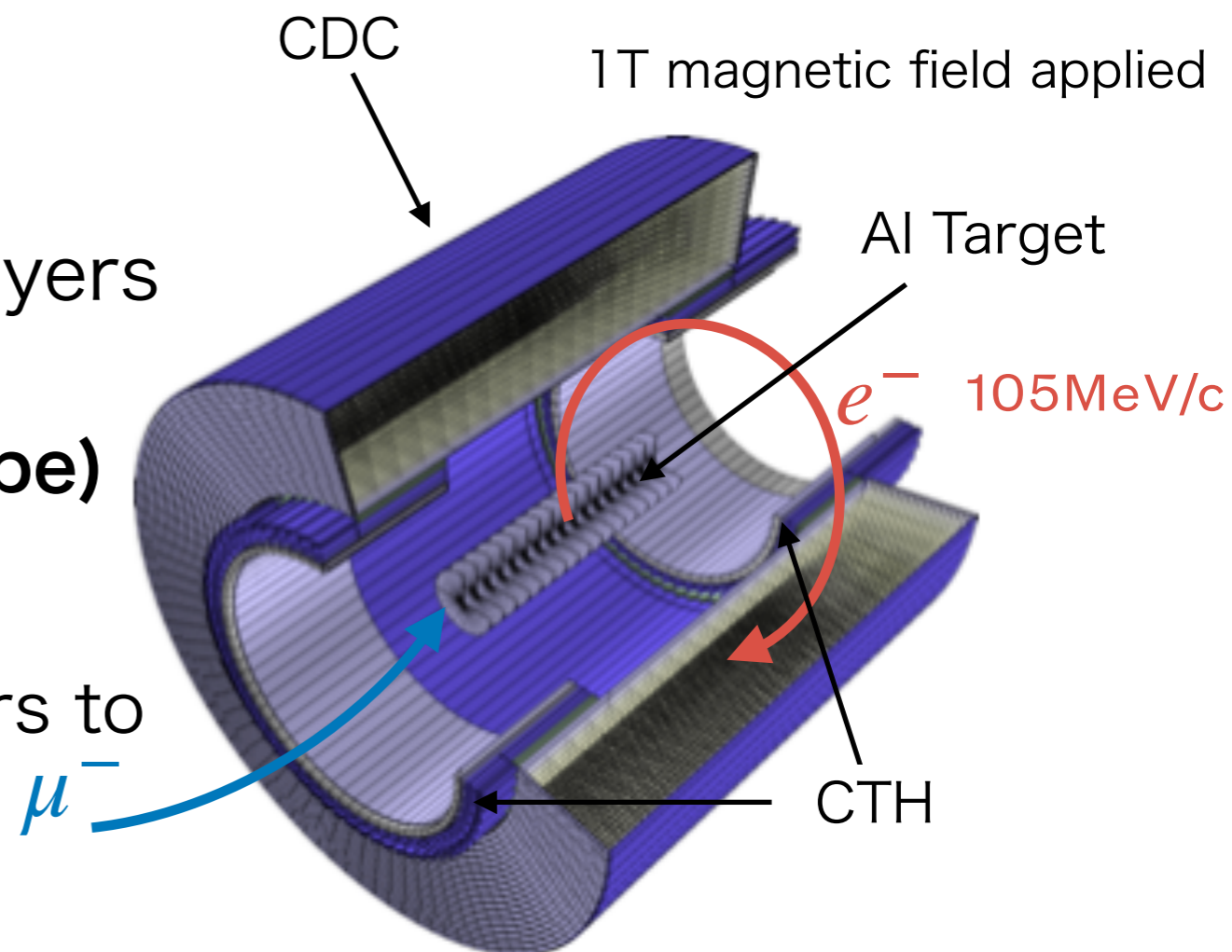
CyDet : Cylindrical detector system

CDC (Cylindrical Drift Chamber)

- Measure the particle momentum
 - 4986 sense wires, 20 stereo layers

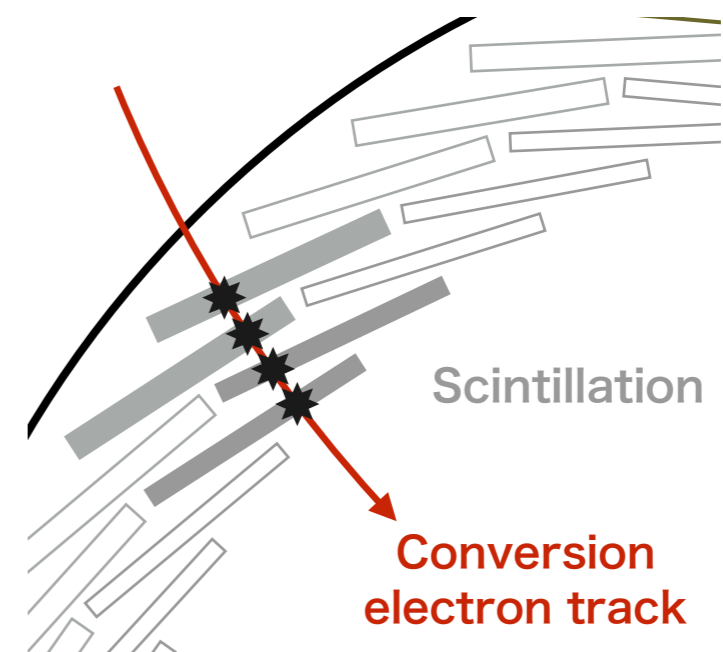
CTH (Cylindrical Trigger Hodoscope)

- Measure the electron timing
- Double layered Scintillation counters to make a primary trigger signal

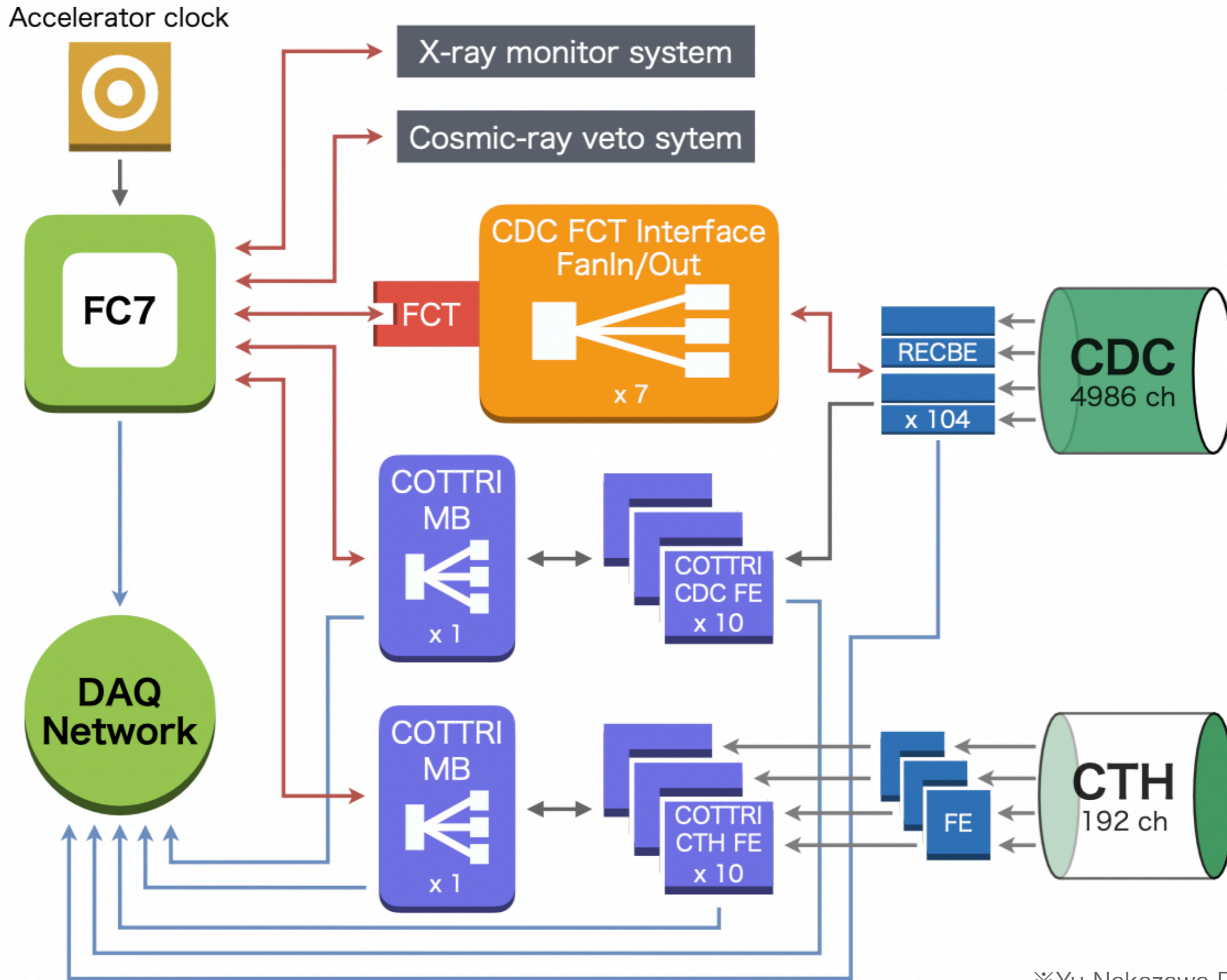


CTH trigger rate ~ 91 kHz (expectation)

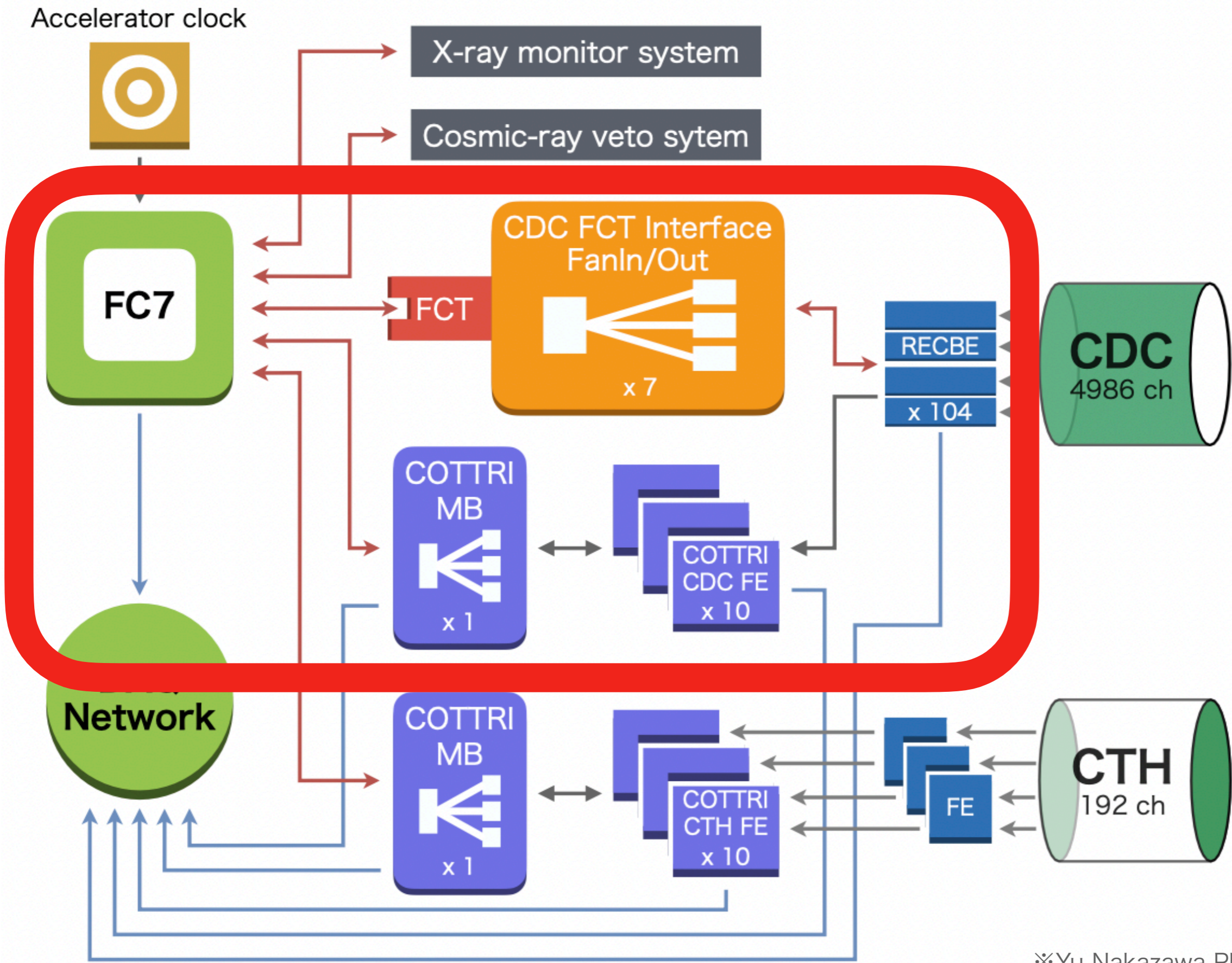
- 4 fold coincidence
- low-E electron dominant



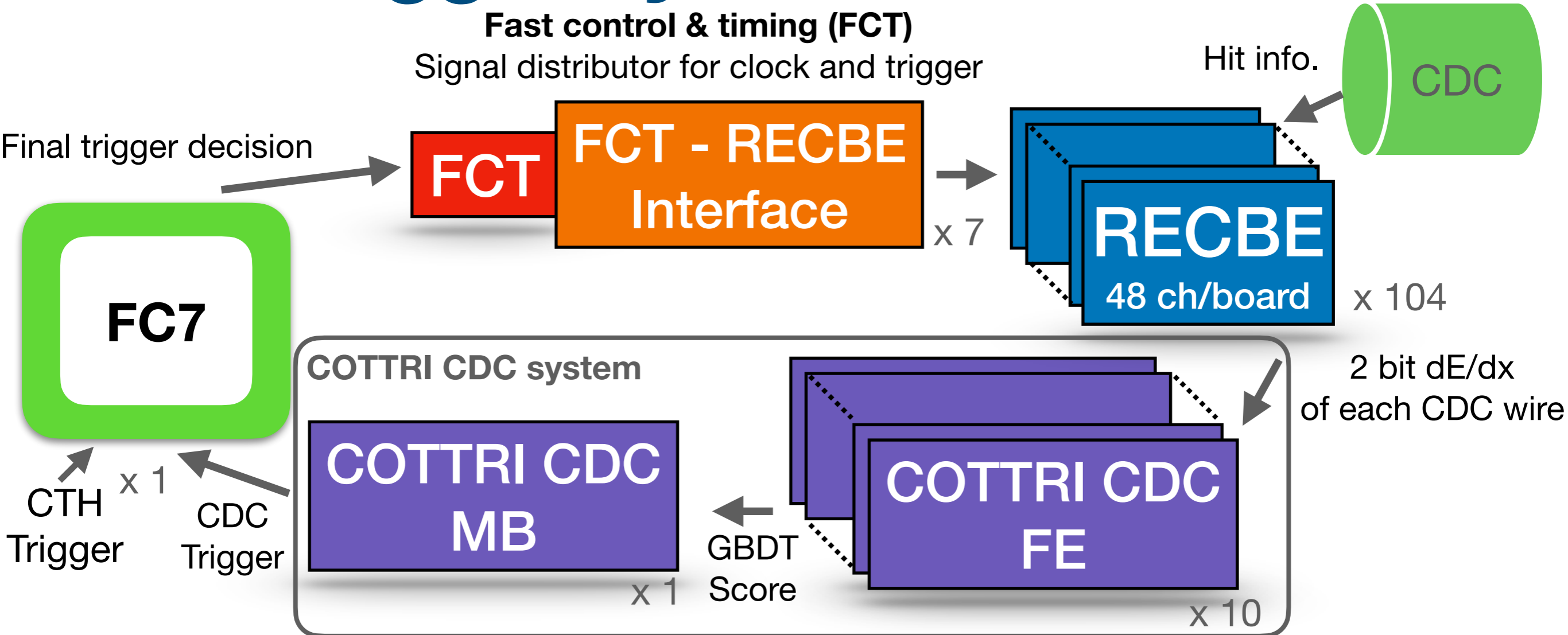
Online trigger system



Online trigger system

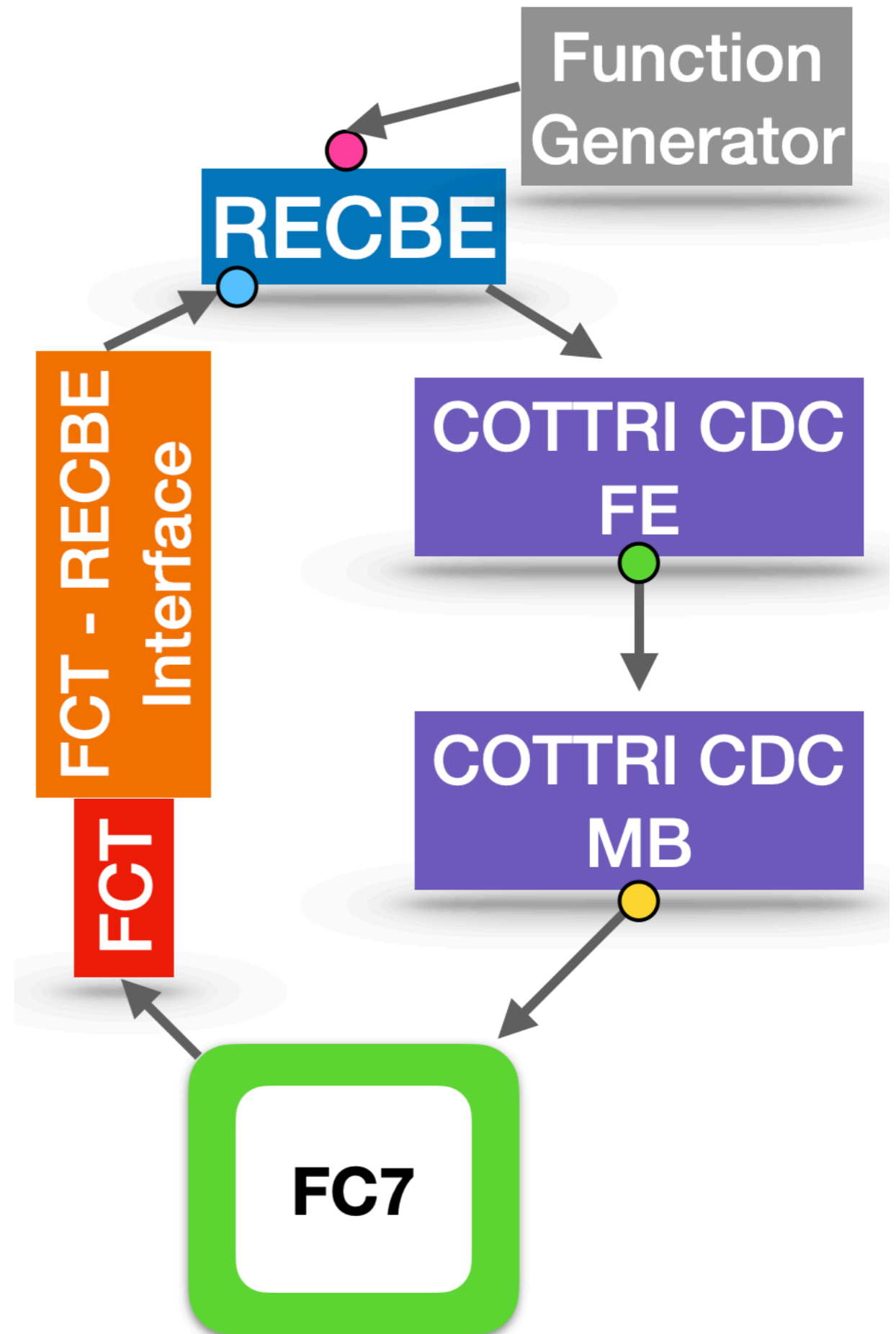
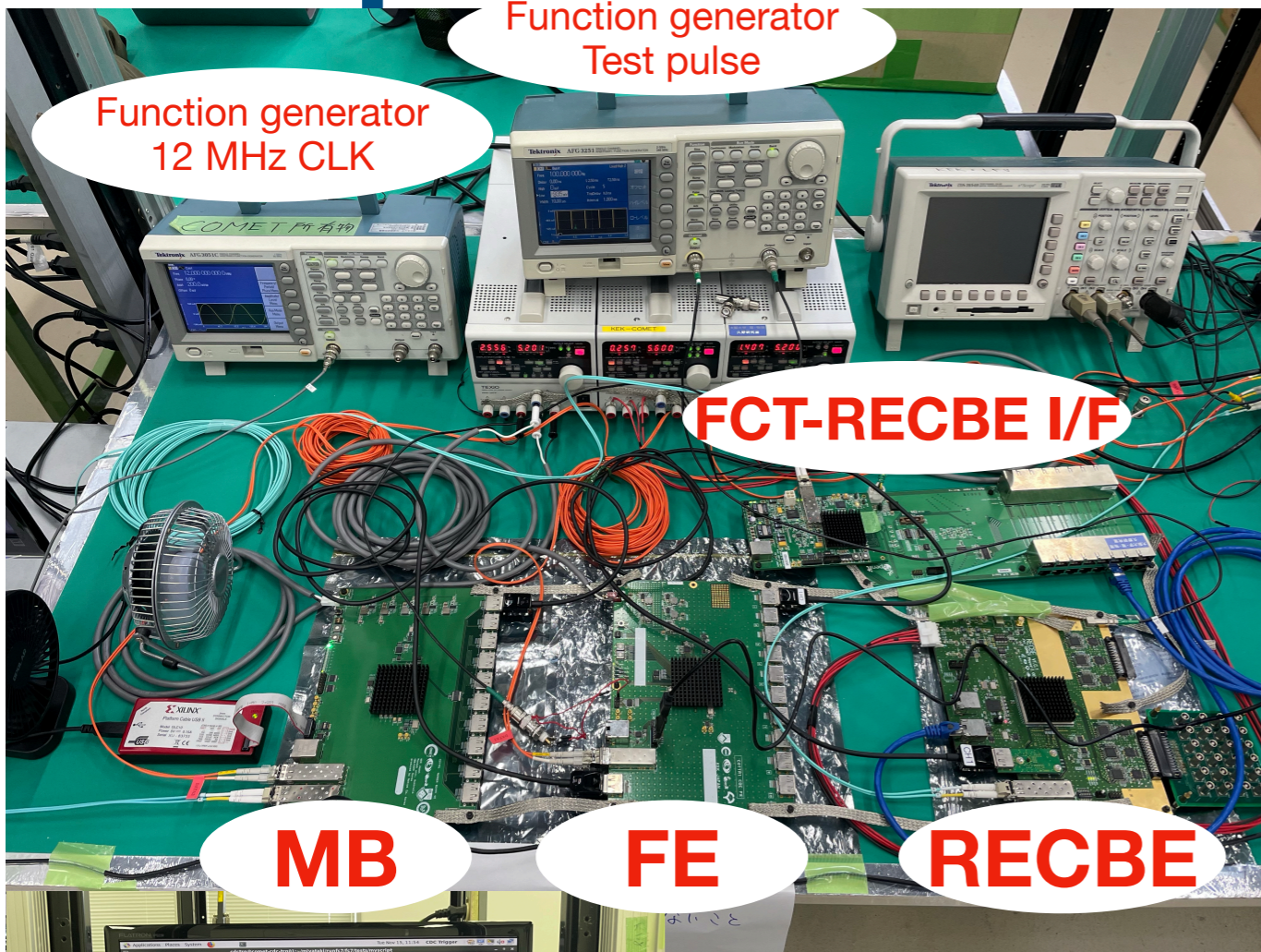


Online trigger system

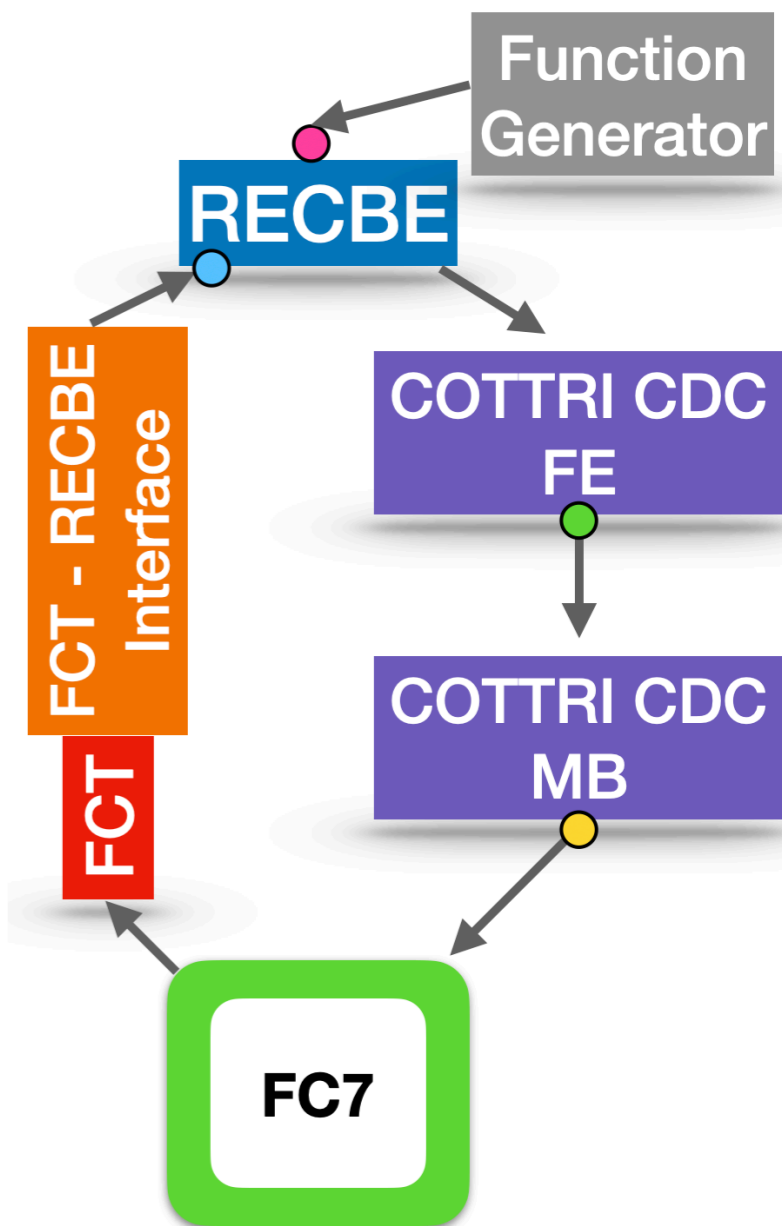
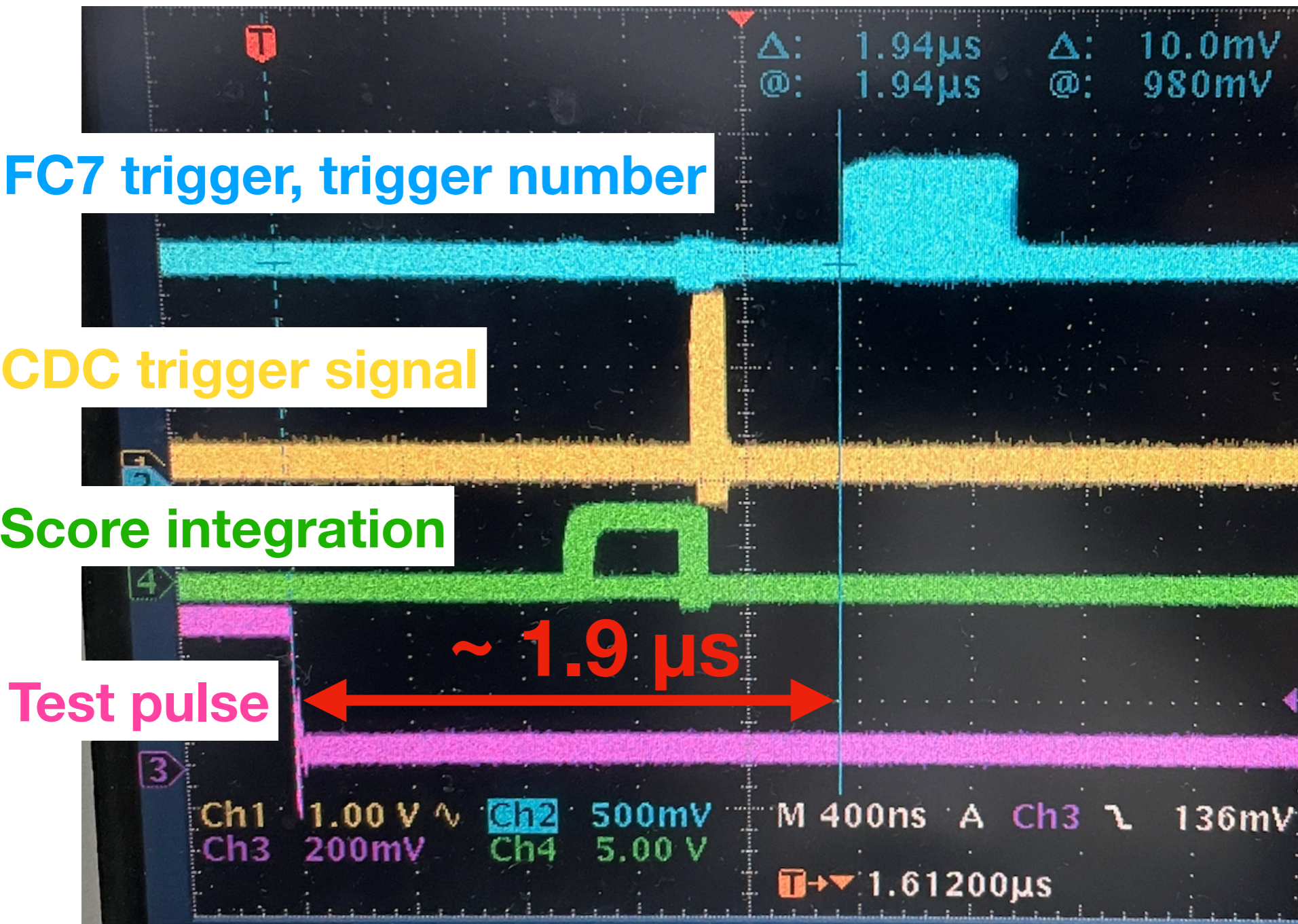


- RECBEs generate the 2 bit dE/dx information and send it @10 MHz
- COTTRI CDC system
 - FE : hit classification based on local/neighboring features.
 - Convert 2 bit data to GBDT scores in 400 ns integration time window
 - MB : event classification with the global feature.
 - Sumsup the GBDT scores and makes the CDC trigger decision @10 MHz

Set up

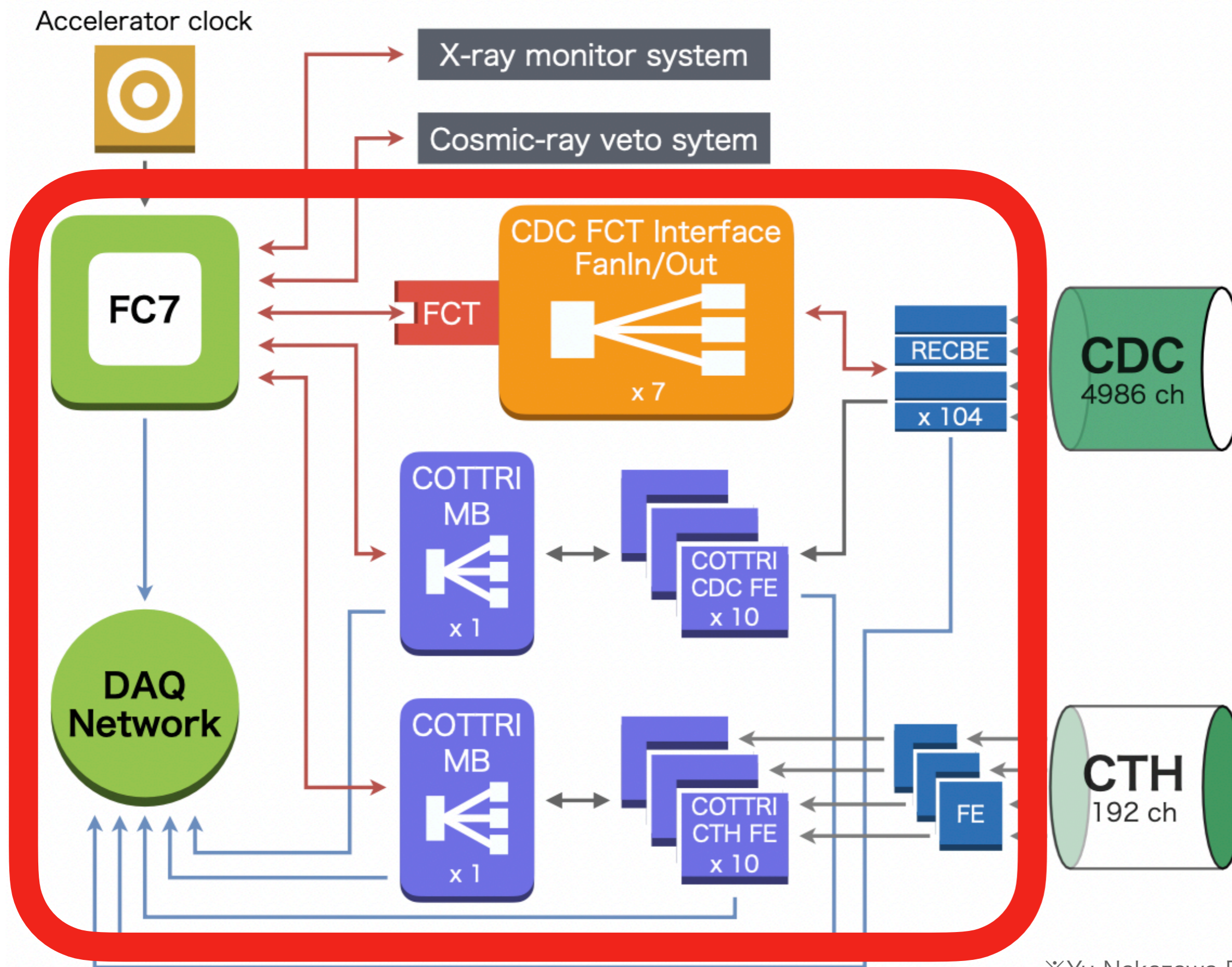


Latency measurement

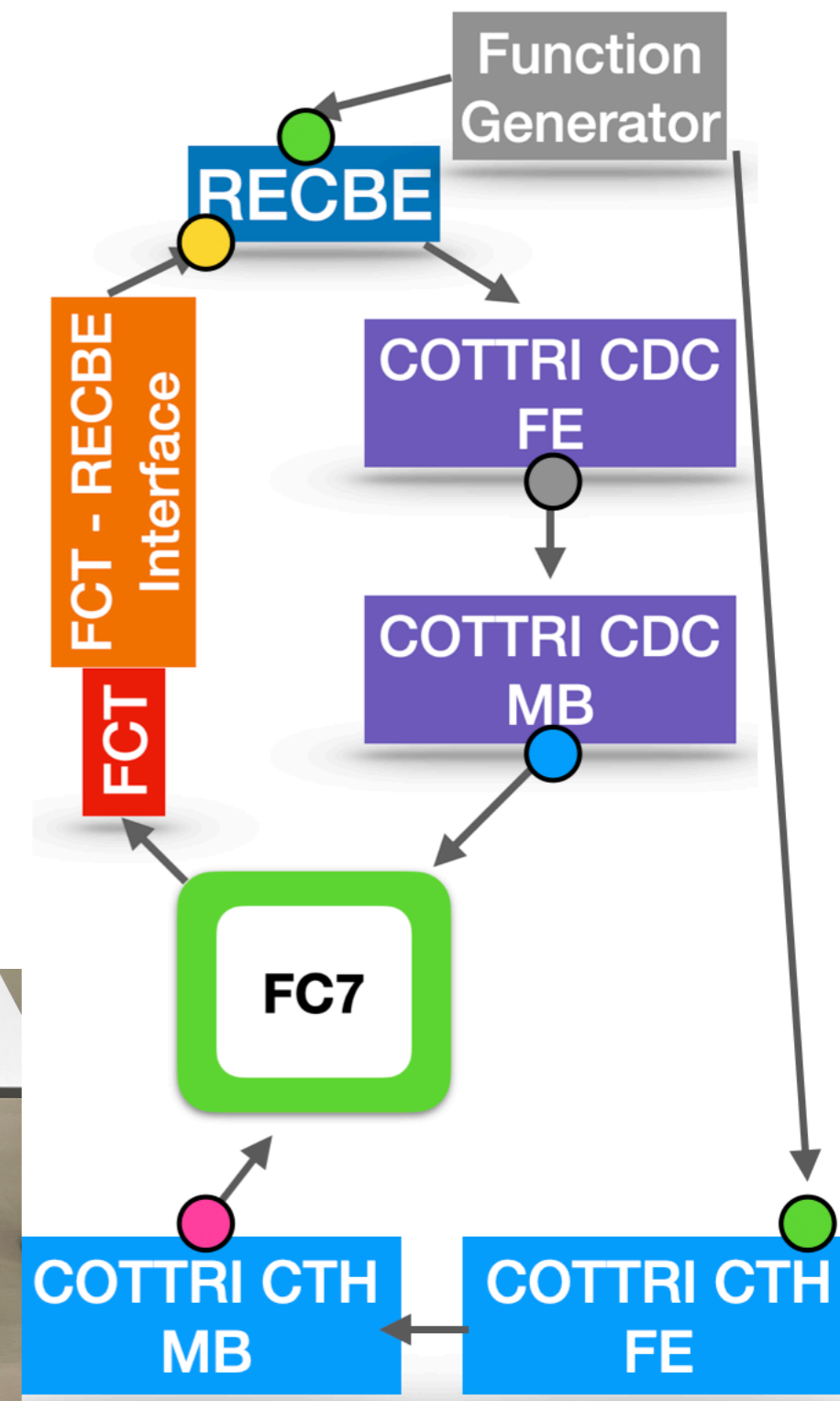
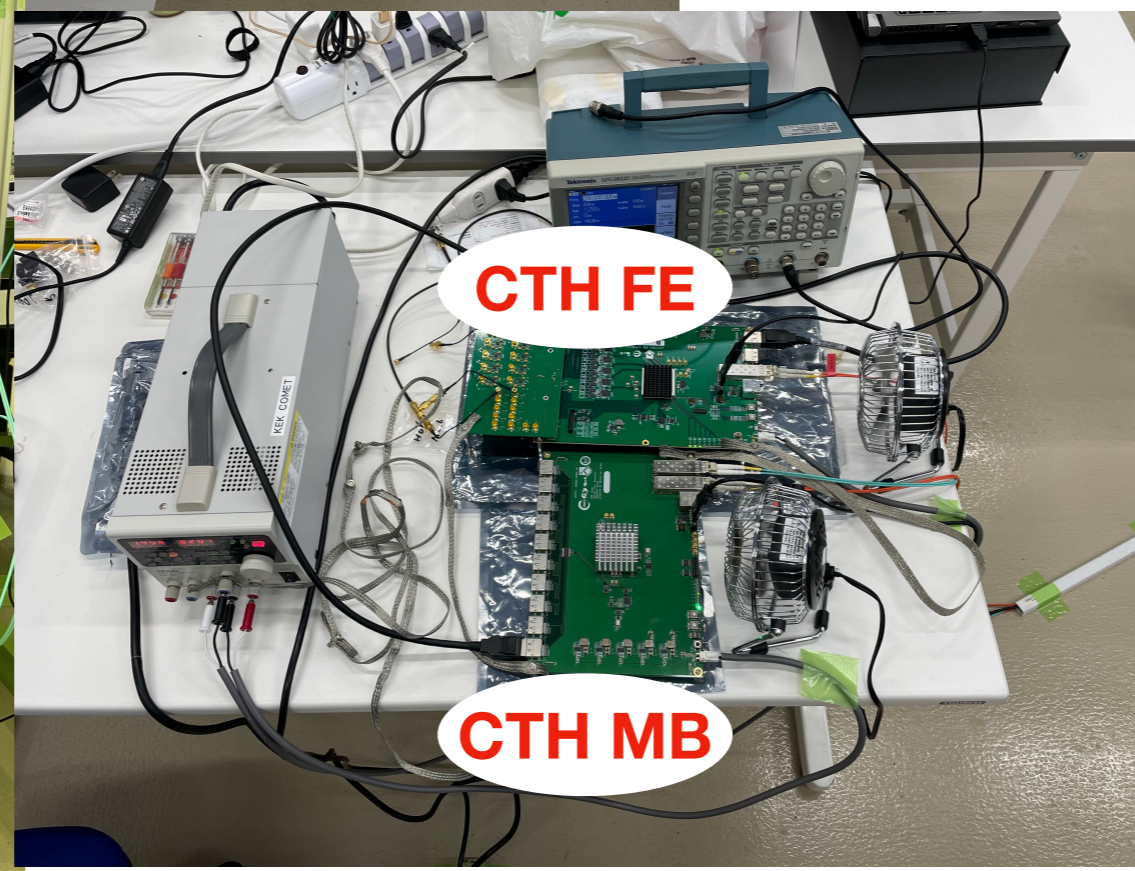
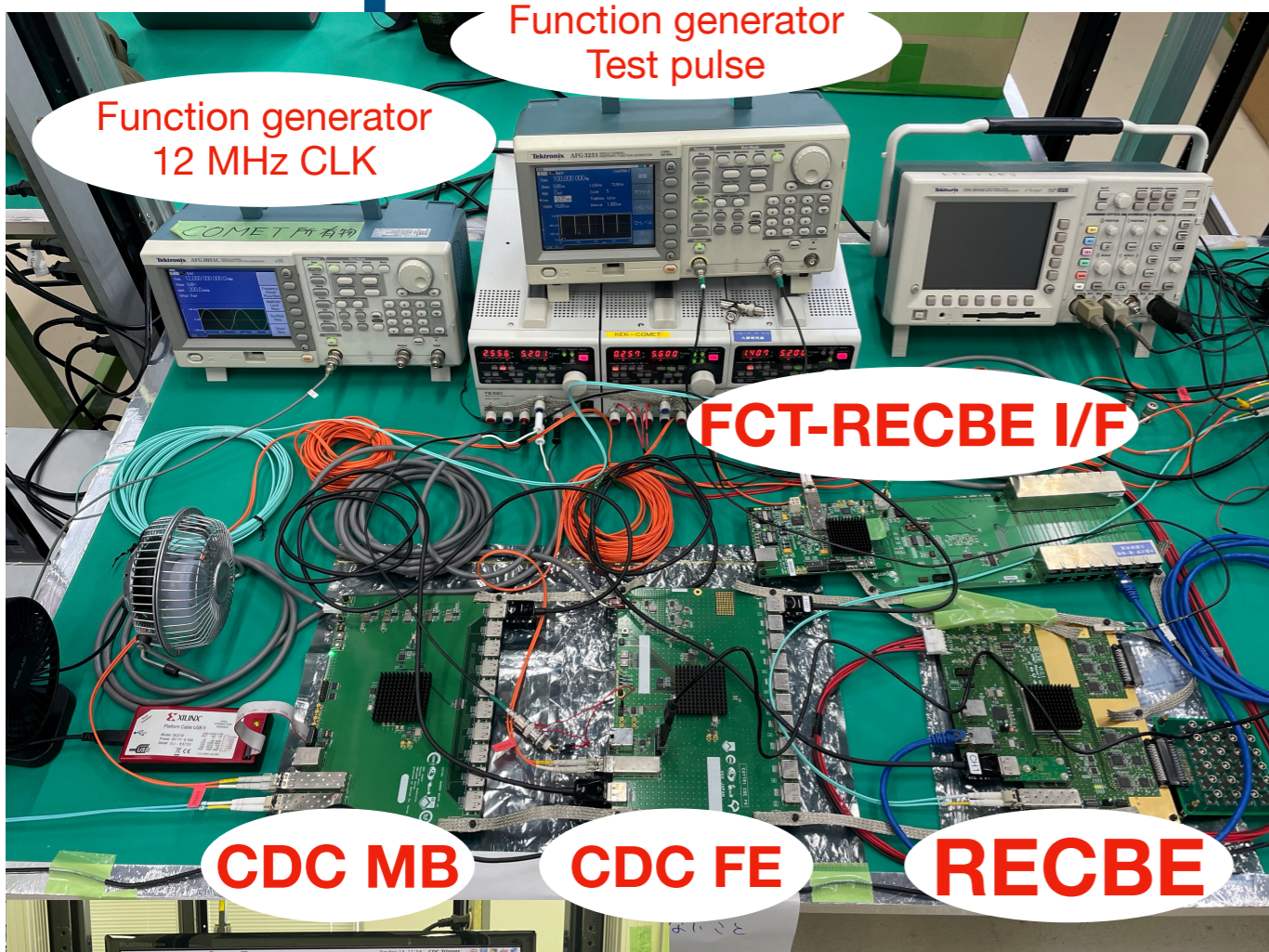


Latency (RECBE - COTTRI CDC system - FC7 - FCT · RECBE I/F - RECBE) is $\sim 1.9 \mu\text{s}$.

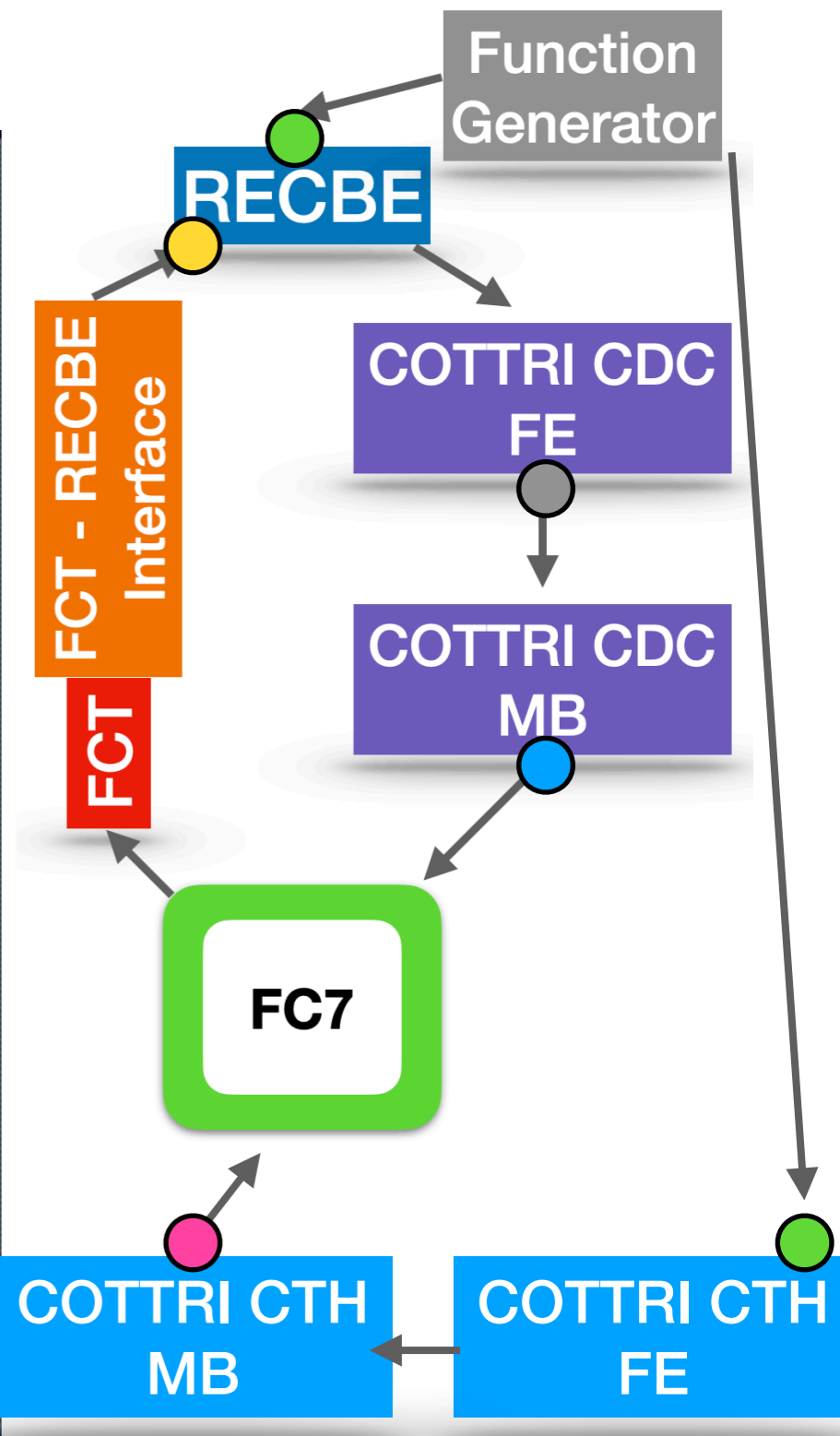
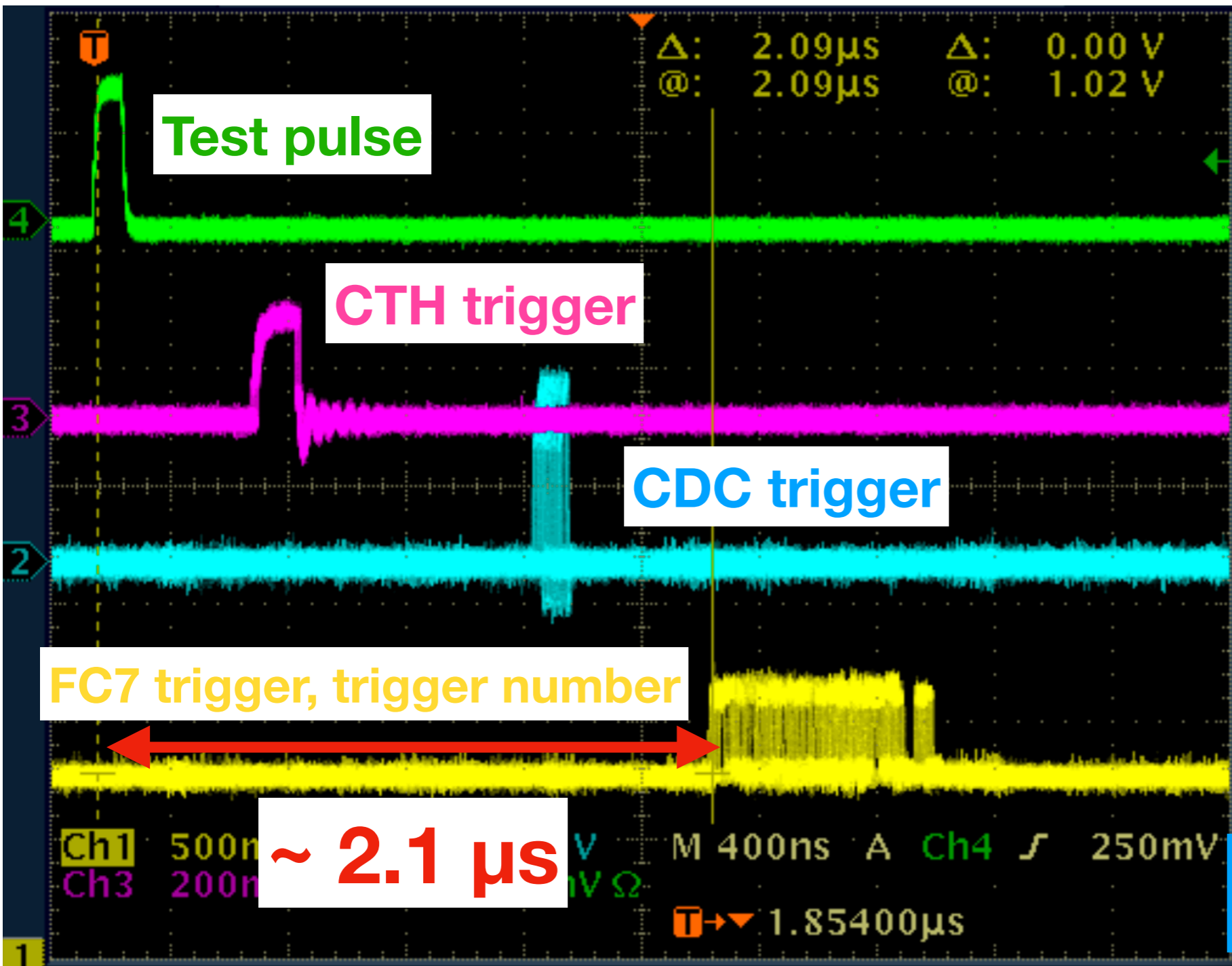
Online trigger system



Set up



Latency measurement

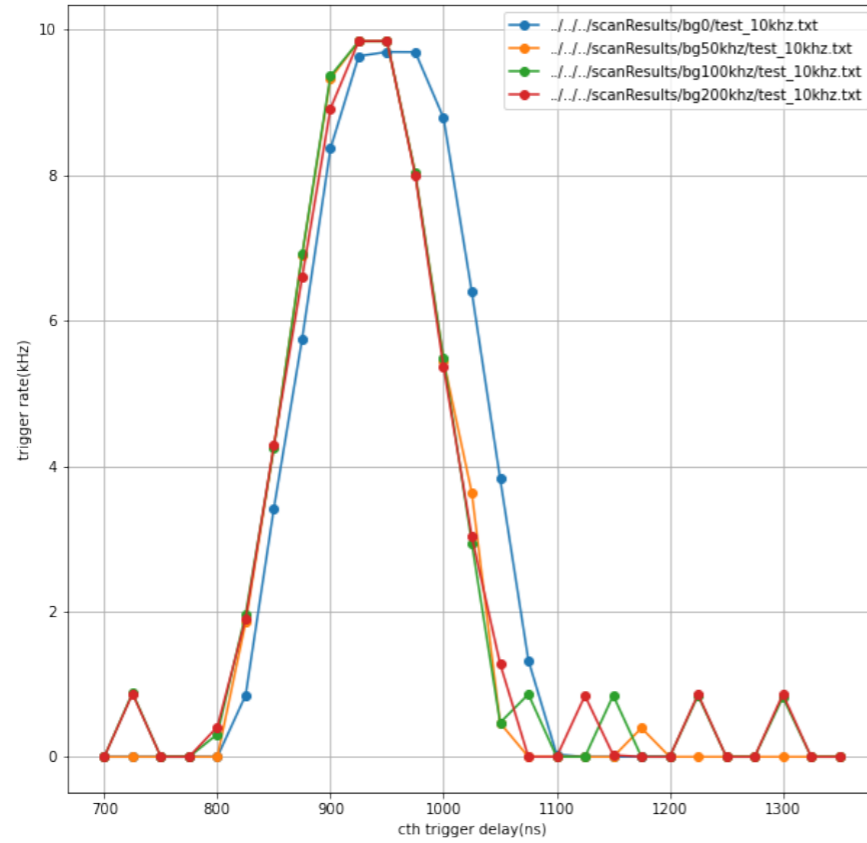


Latency $\sim 2.1 \mu\text{s}$

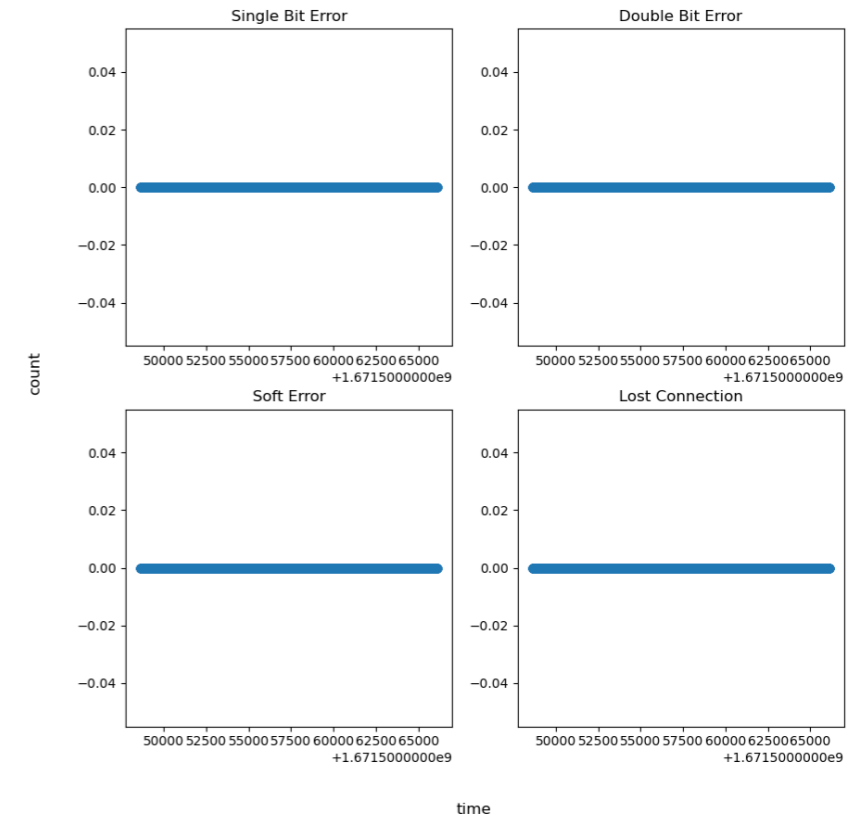
There is no increase in delay due to the coincidence trigger.

Other preliminary results of FC7 coincidence trigger

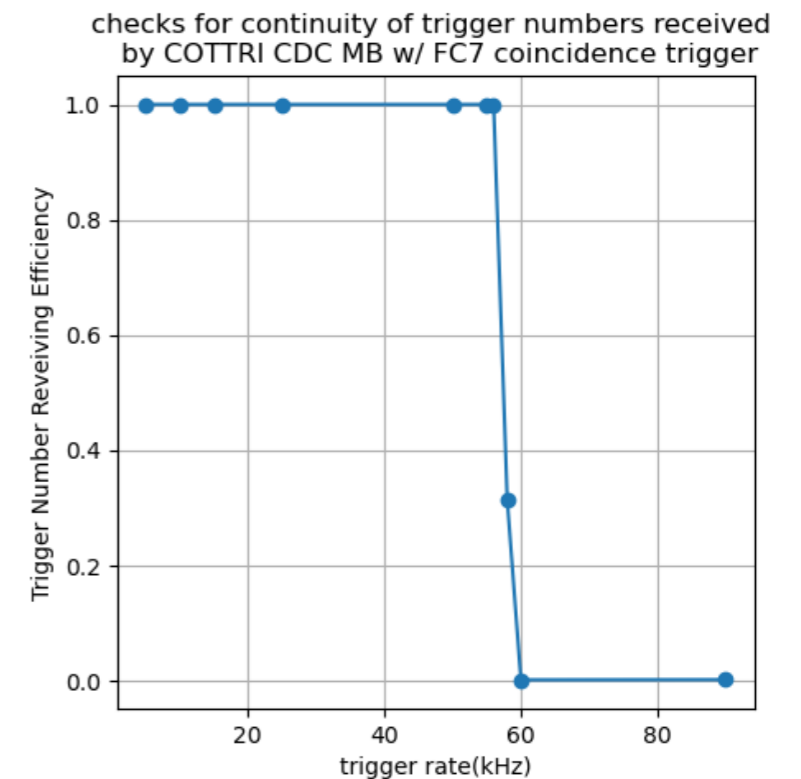
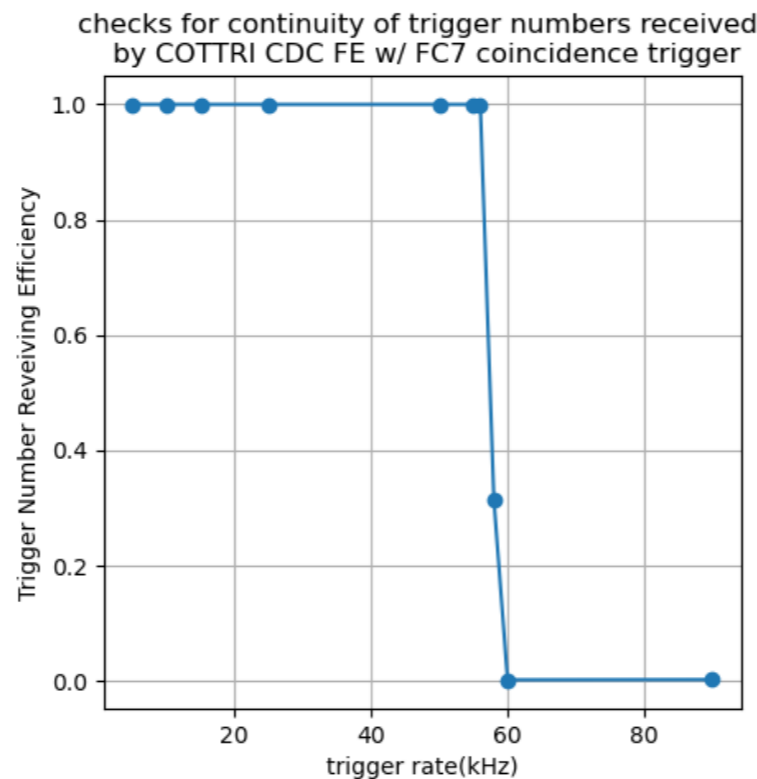
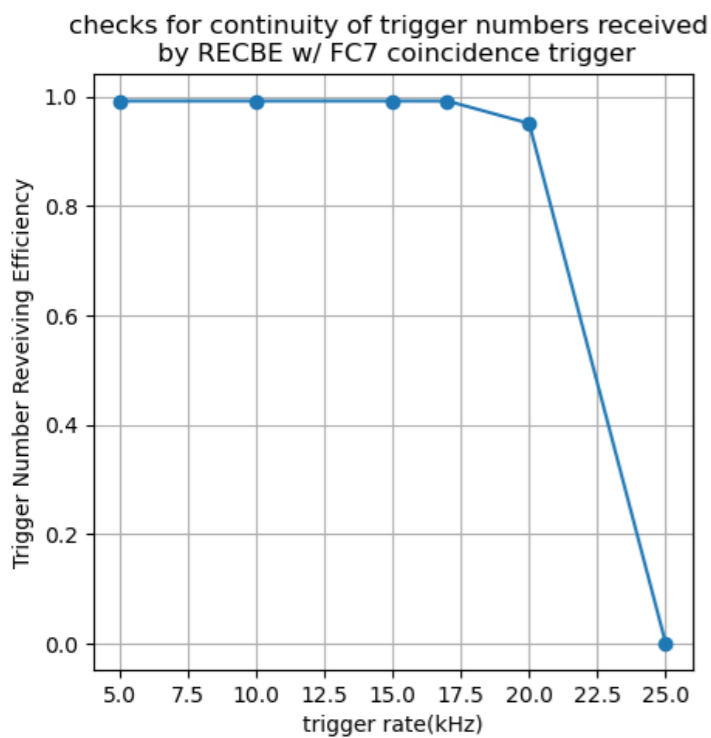
coincidence delay scan



Communication stability check



Trigger number continuity check results



To improve signal acceptance, increase trigger efficiency, and widen the measurement time window

Measurement time window(ns)	Current system [700,1170]	[500,1170]
Signal acceptance(%)	4.2	7.0

x 1.7

The current trigger system has a trigger rate above 13 kHz for 500 ns start

The current CDC trigger algorithm

= hit classification by ML + event classification by counting signal-like hits



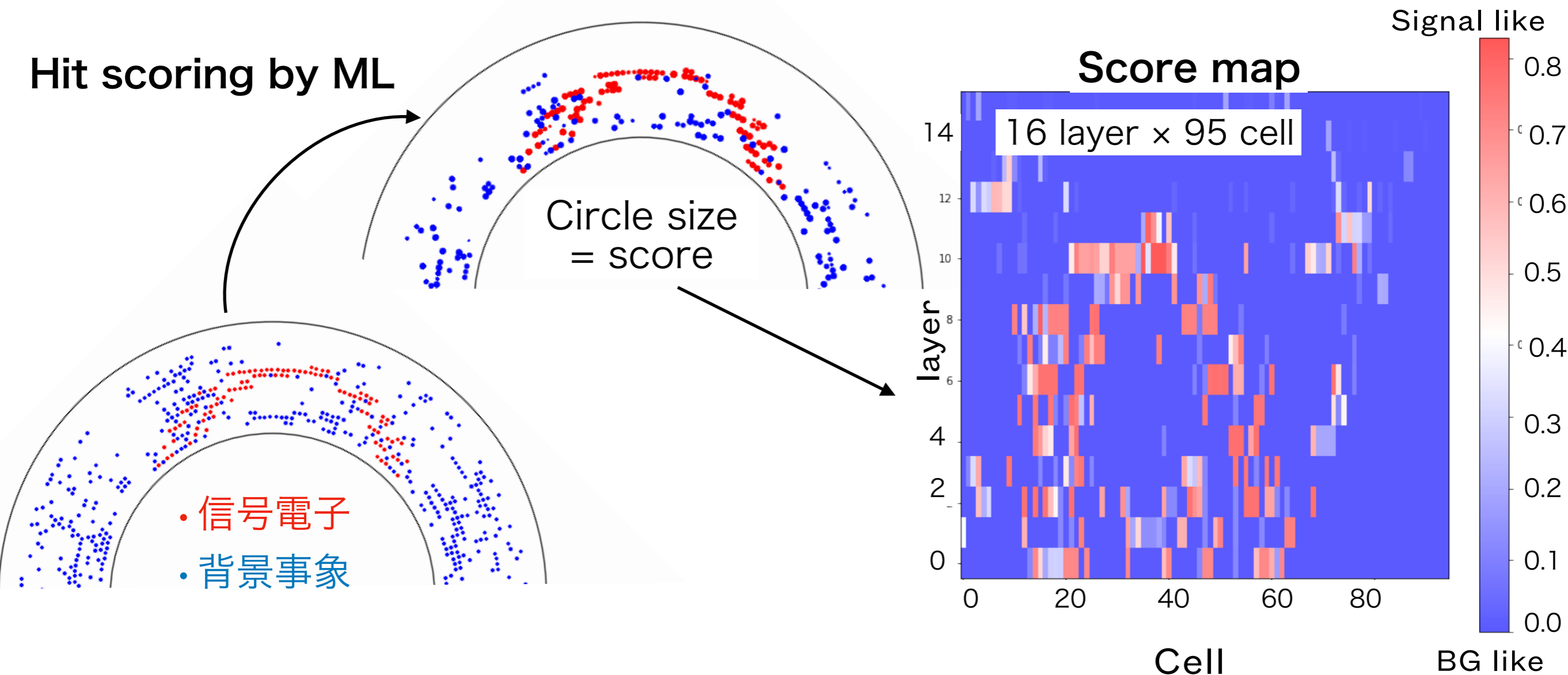
New CDC trigger algorithm under development

= hit classification by ML + **event classification by Neural Network**

Neural network based event classification 15

New CDC trigger algorithm under development

= hit classification by ML + event classification by Neural Network

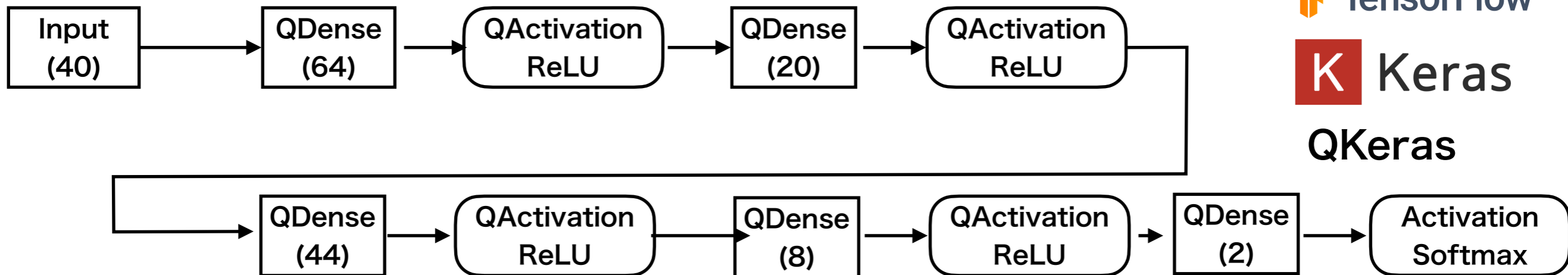


- By using the score map as input to the Neural network, **pattern recognition of the trajectory drawn by the signal electrons** can be introduced for event classification.

Model construction

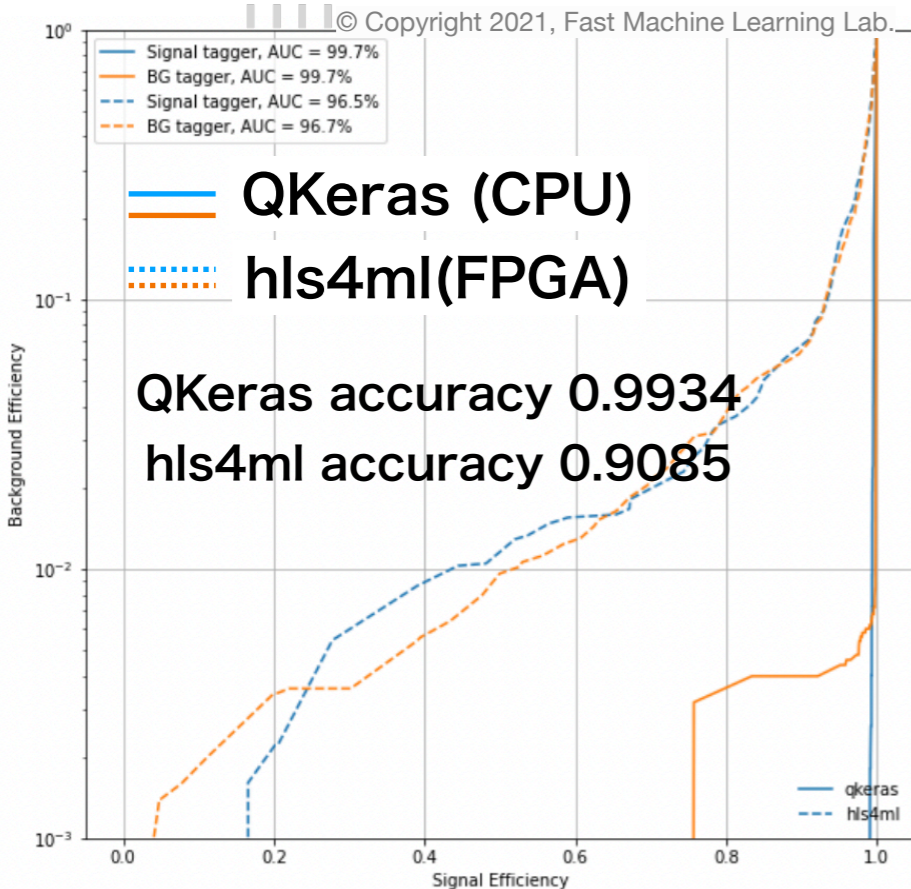
Fully connected 5 dense layer Quantized MLP

(HPs tuned except for bit width)



 TensorFlow
 Keras
QKeras

hls4ml



Convert models into firmware-transformable code(RTL) w/o programming in HDL

<https://dx.doi.org/10.1088/1748-0221/13/07/P07027>

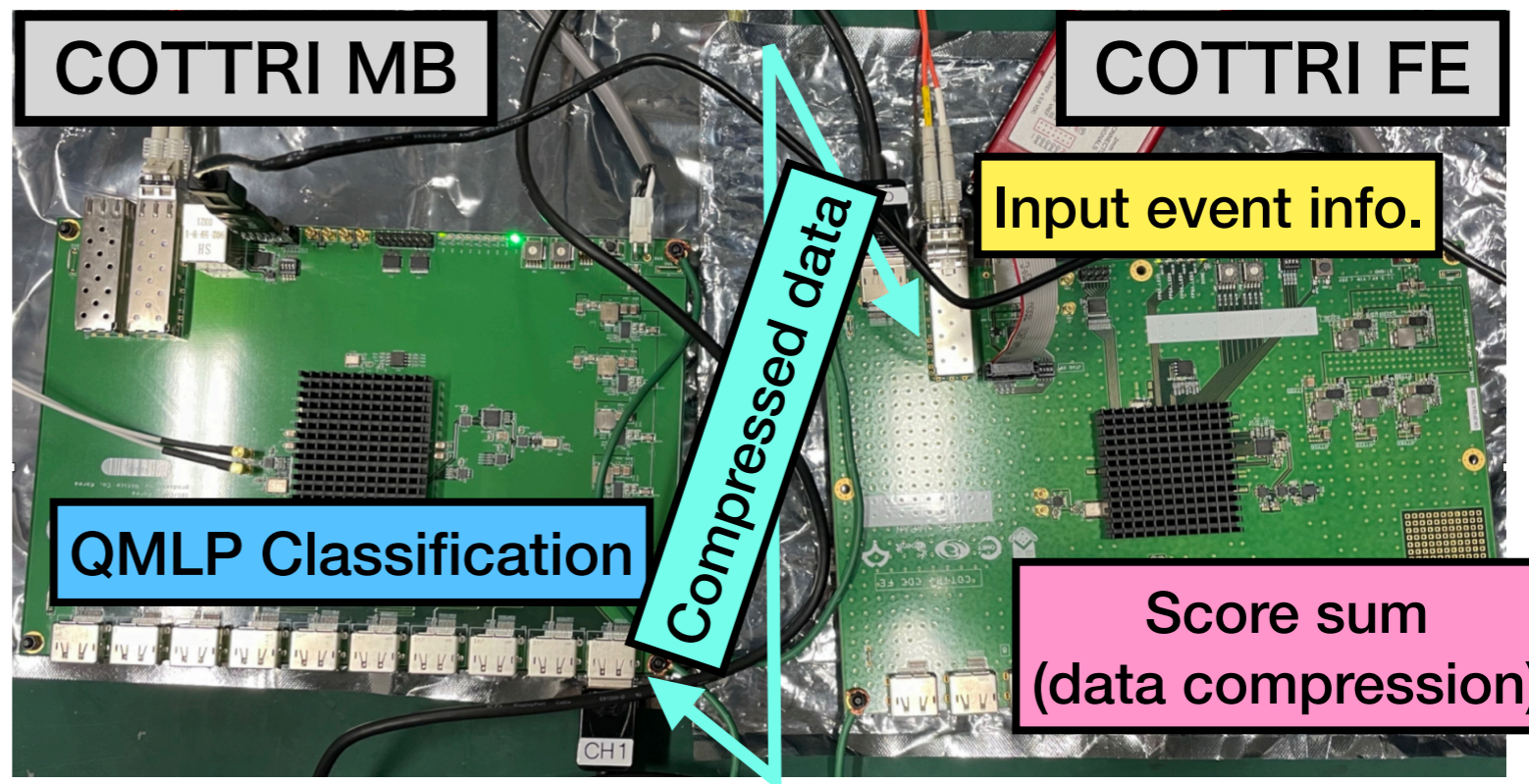
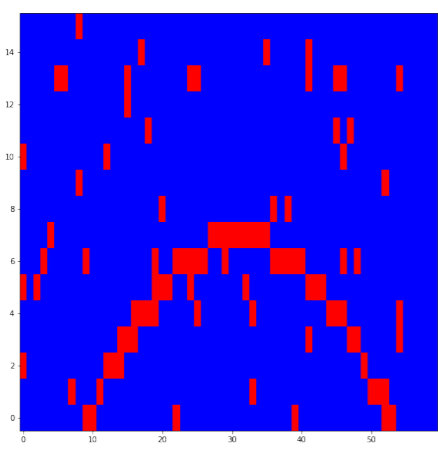
FPGA : AMD Xilinx Kintex-7 xck355t-ffg901-1

	Usage (%)			
Latency @200MHz	BRAM	DSP	FF	LUT
130 ns	~0	~0	5	32

After high-level synthesis of the C++ file generated by hls4ml with vivado_hls, I generated this QMLP ip in vivado and implemented it into COTTRI MB's firmware.

Test

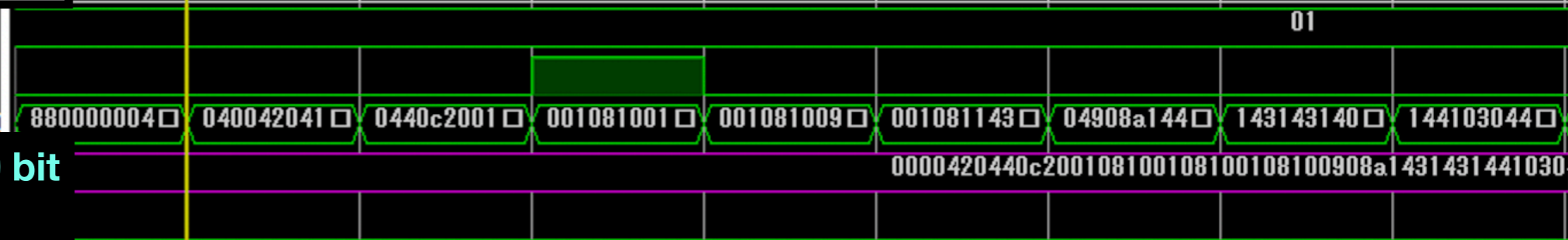
Input event



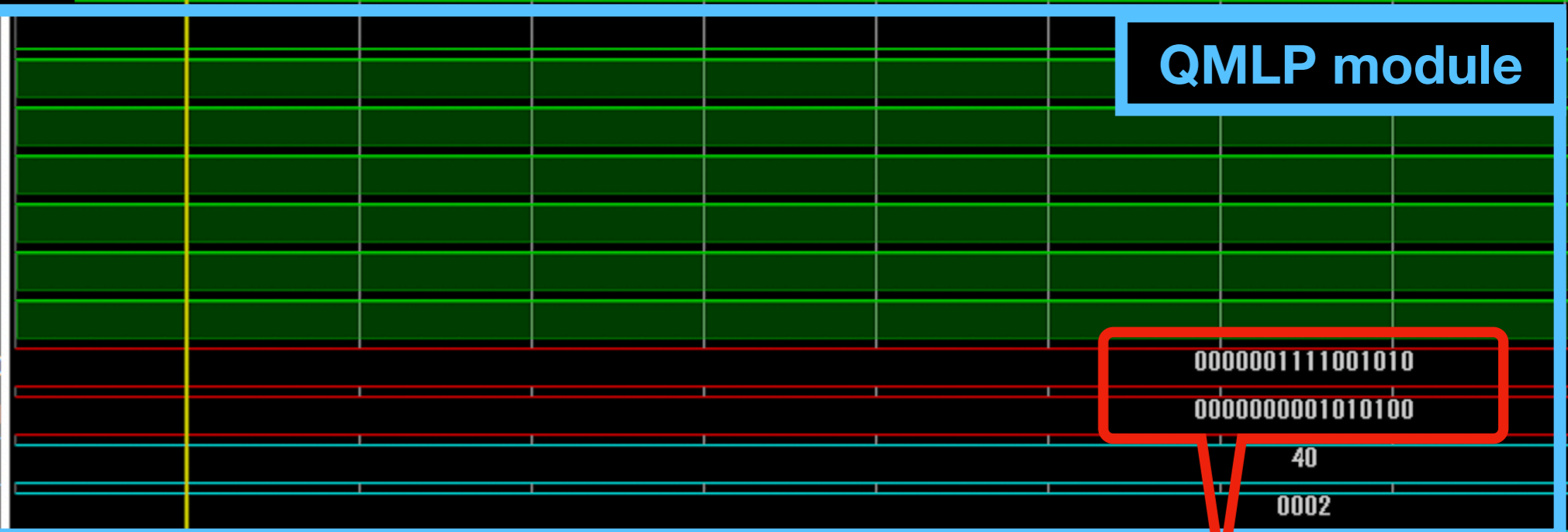
ILA : COTTRI MB

- > NUM_OF_VALID_FE[7:0]
- └ COTTRI_VALID
- > DpRxDataOut[239:0]

Compressed data 240 bit from COTTRI FE



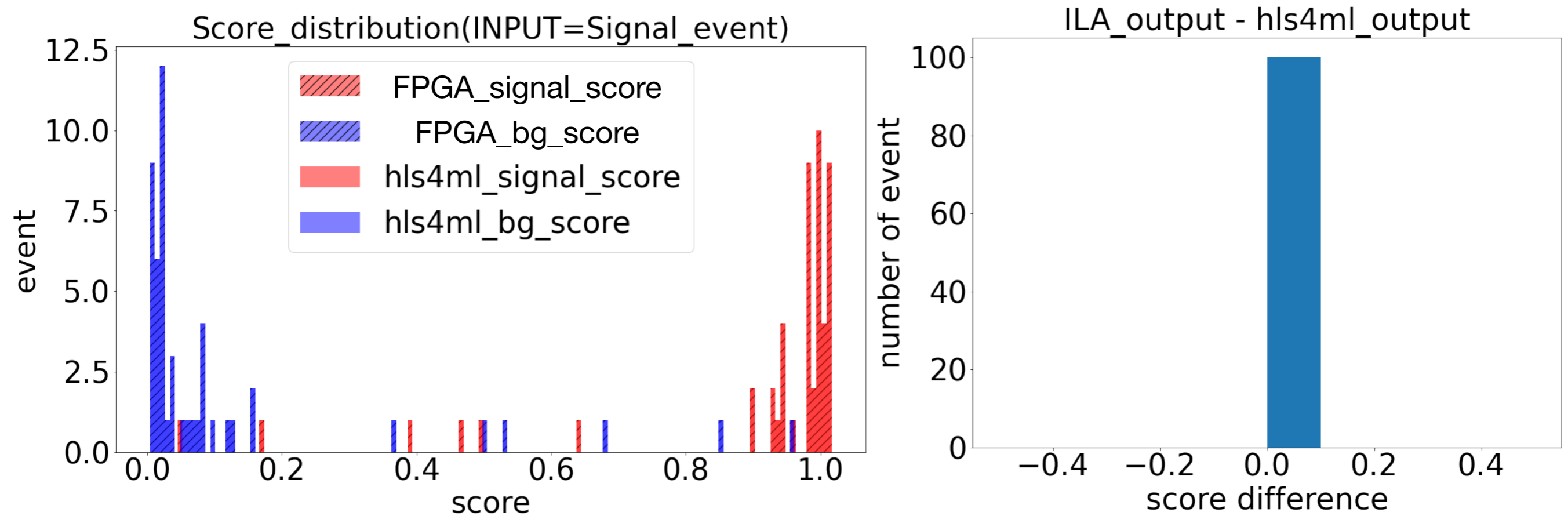
- └ MLP_IDLE
- └ MLP_READY
- └ MLP_DONE
- └ SIGNAL_OUT_VALID
- └ BG_OUT_VALID
- └ CONST_SIZE_IN_1_VALID
- └ CONST_SIZE_OUT_1_VALID
- > **Signal score**
- > **BG score**
- > CONST_SIZE_IN_1[15:0]
- > CONST_SIZE_OUT_1[15:0]



BG score < **Signal score**
 0.08 < 0.95

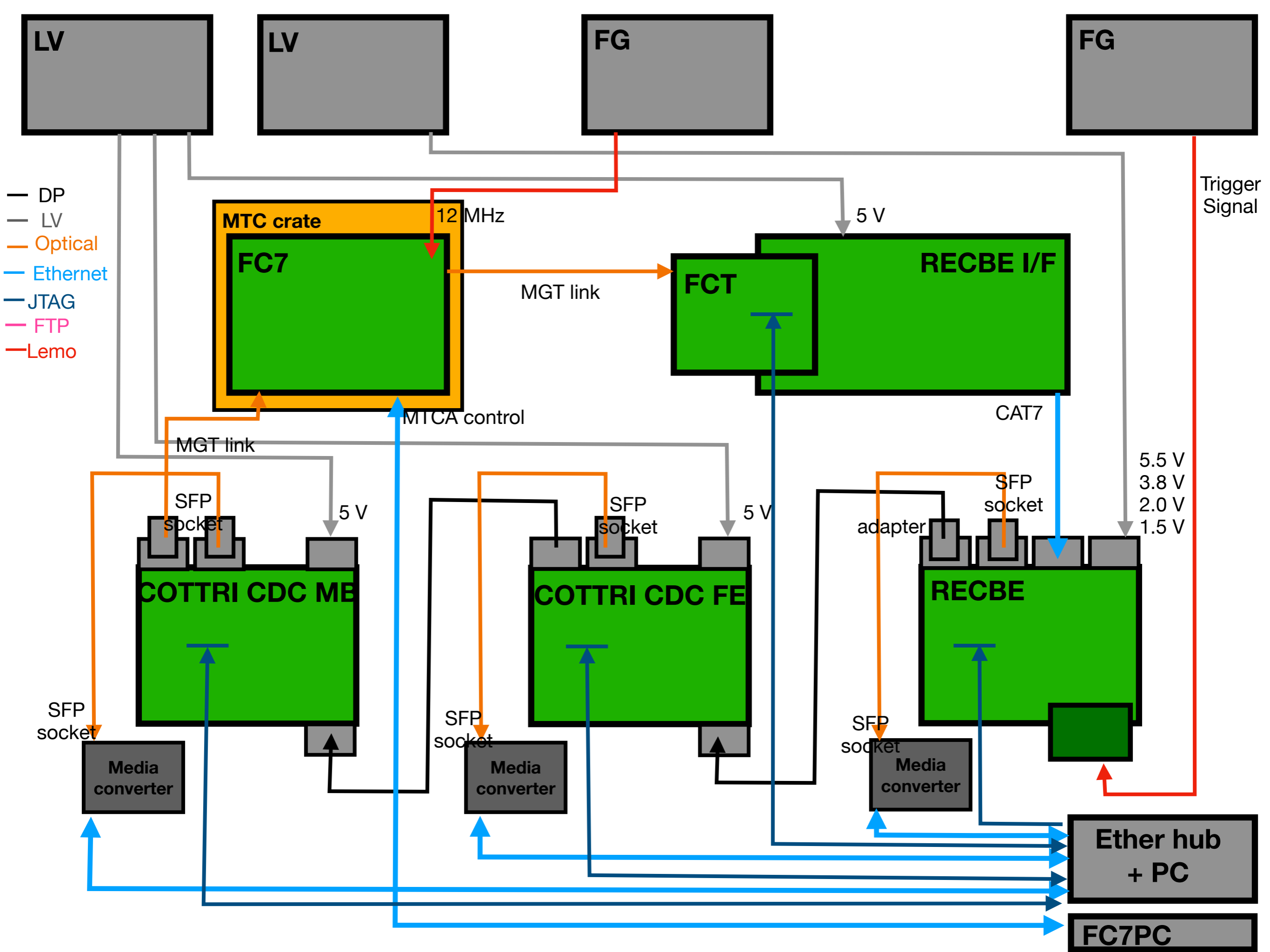
Score as expected 😎

Neural Network test results



software and hardware predicted values are in perfect agreement

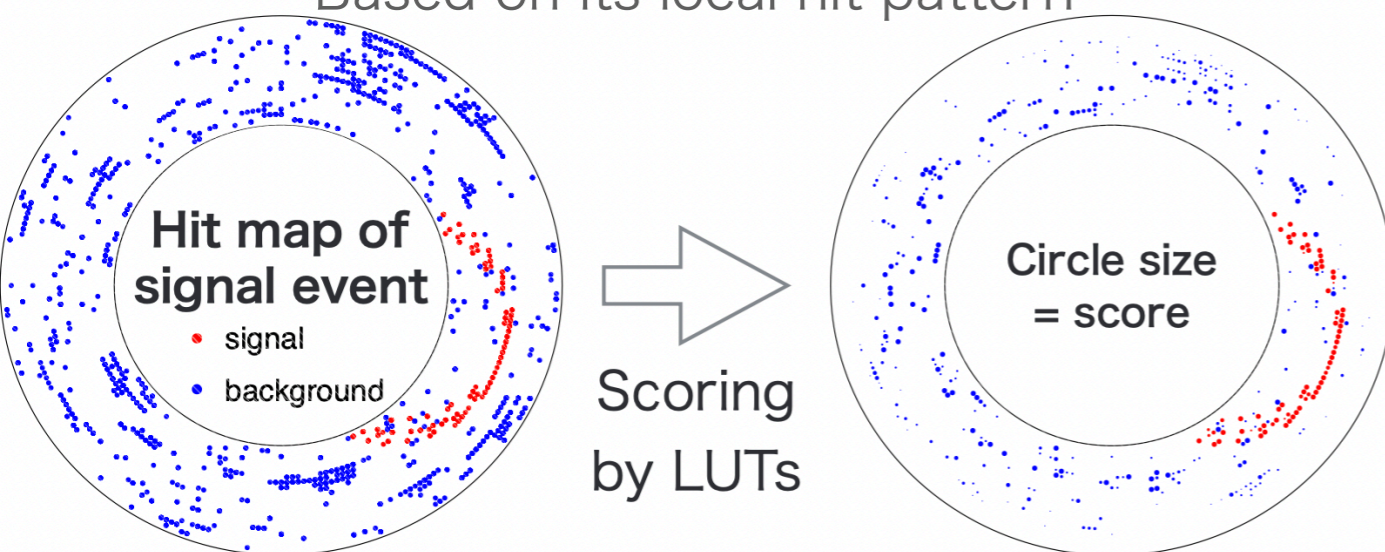
- CDC trigger chain test was conducted.
 - Latency $\sim 2 \mu s$
- CyDet trigger chain test was conducted.
 - The FC7 trigger was successfully issued and distributed to whole CyDet trigger system.
 - Latency $\sim 2 \mu s$
 - Many tests were carried out. The analysis is ongoing.
- New online event classification study is ongoing.
 - Neural network model that classifies the signal electron event and background event was constructed.
 - The trained model was successfully implemented on FPGA.
 - Software and hardware prediction scores are in perfect agreement.



Trigger algorithm

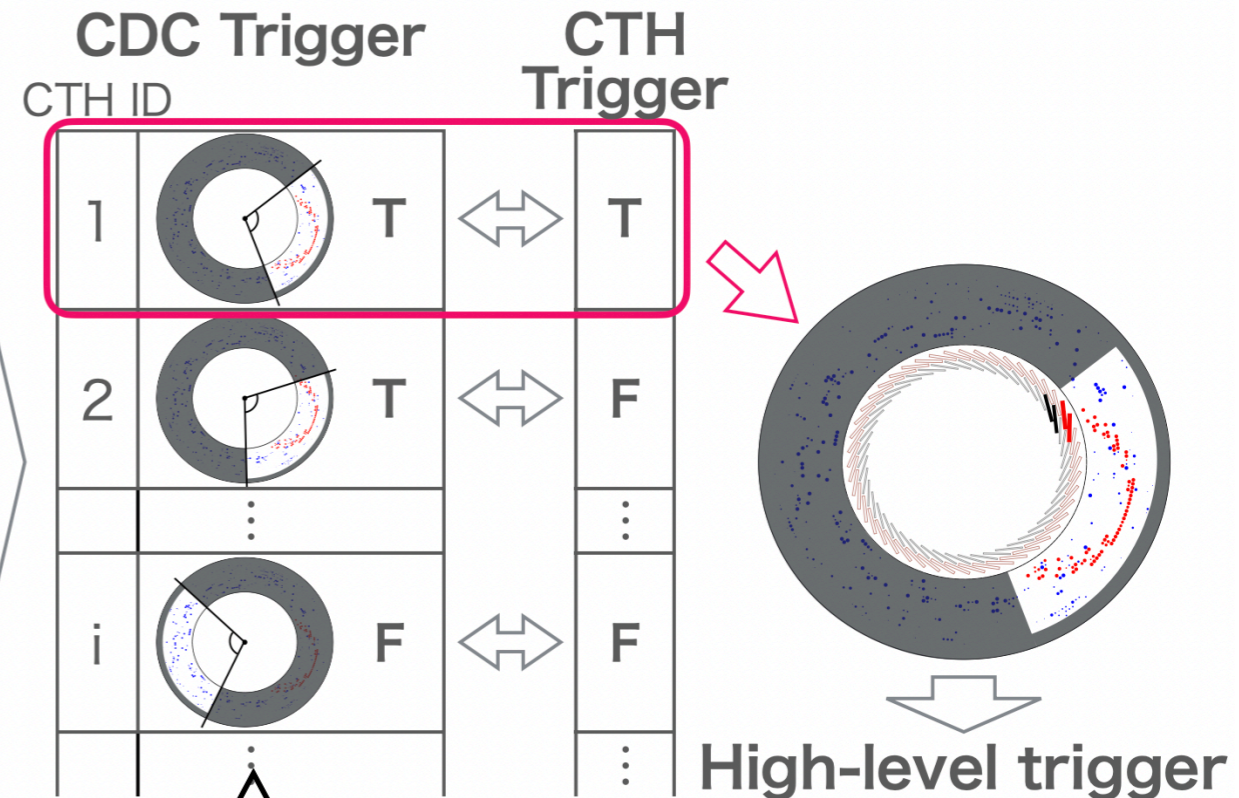
Hit classification

GBDT gives each wire hit a score
Based on its local hit pattern



Yu nakazawa PhD thesis Fig3.5

Event classification



The current CDC trigger algorithm

1. Set the CDC active section for each segment of CTH
2. Within each active section, count hits that exceed the score threshold
3. CDC trigger is issued when the count exceeds the threshold.

New CDC trigger algorithm under development

1. Set the CDC active section for each segment of CTH
- 2'. Execute Neural Network inference w/ score information of each active section as inputs
- 3'. CDC Trigger is issued based on Neural Network classification.

Signal and BG hits

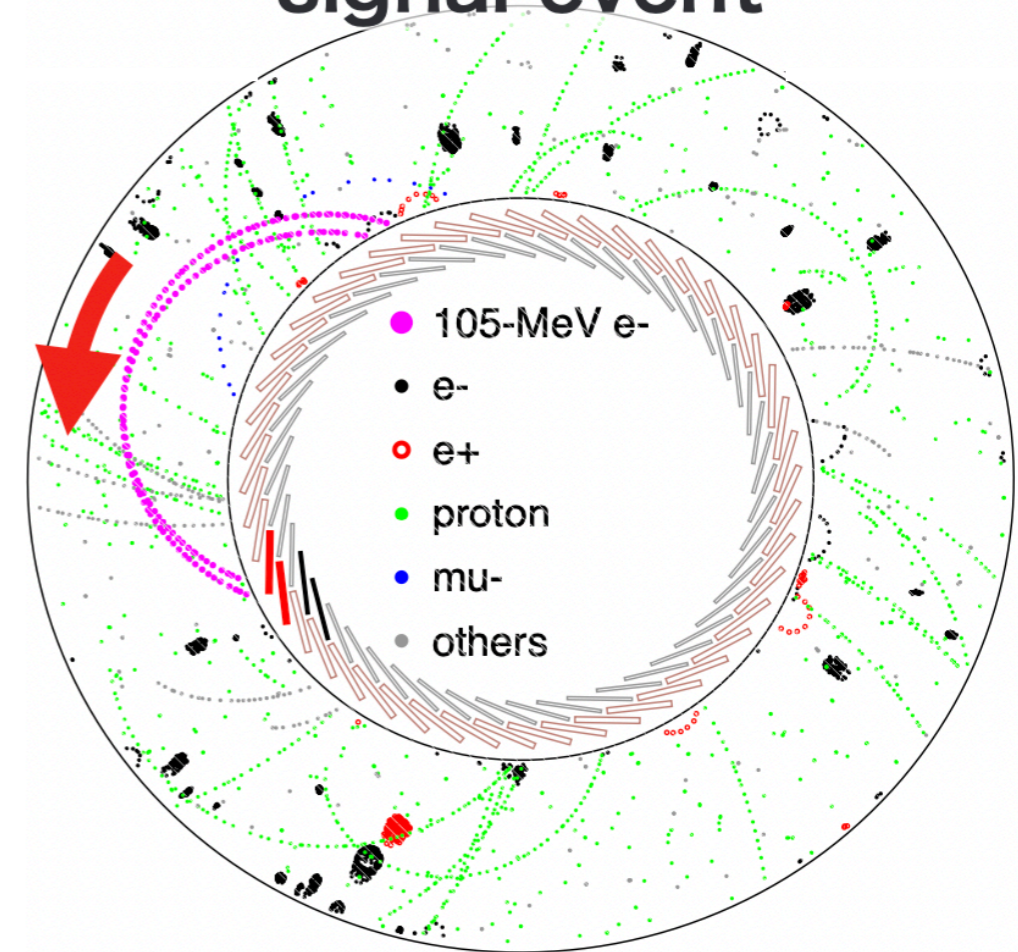
Signal-hit characteristics

- **Contained helical tracks**
- **Single hit in the same wire**
- **MIP-level energy loss**

Background-hit characteristics

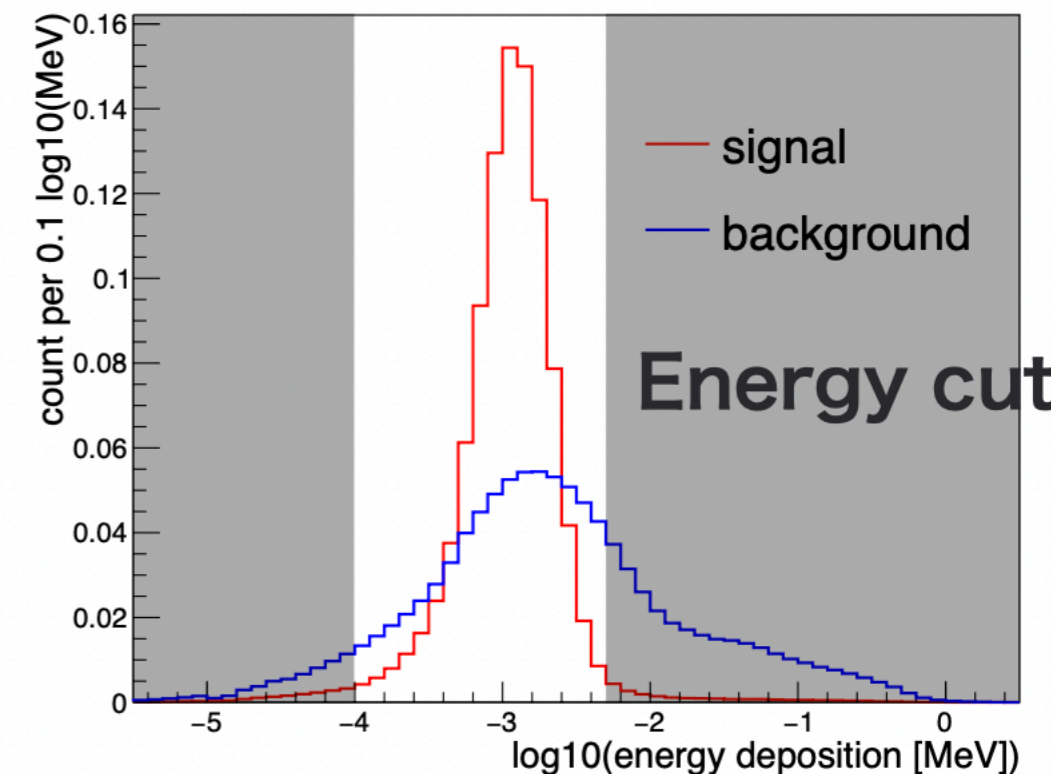
- Low energy electrons
 - Interaction of gamma rays at the CDC walls
 - Helical trajectory contained in the same cell
 - **Multi hits in the same wire**
- Protons (from muon nuclear capture)
 - Momentum higher than 100MeV/c
 - **Large energy loss**
 - ~40 protons/beam-pulse

Hit map of signal event



✧ Yu Nakazawa PhD Thesis Fig3.2

Energy deposition



Hit classification

*GBDT = Gradient Boosted Decision Tree

- Machine learning algorithm (GBDT*) to score hit information for each wire based on energy loss and local patterns

GBDT input features

Interest wire 2bit Edep

count

ADC sum compressed into 2bits

— signal

— Background

LayerID

fraction of CDC hits

layerID

— signal

— Background

Neighboring wire 2bit Edep

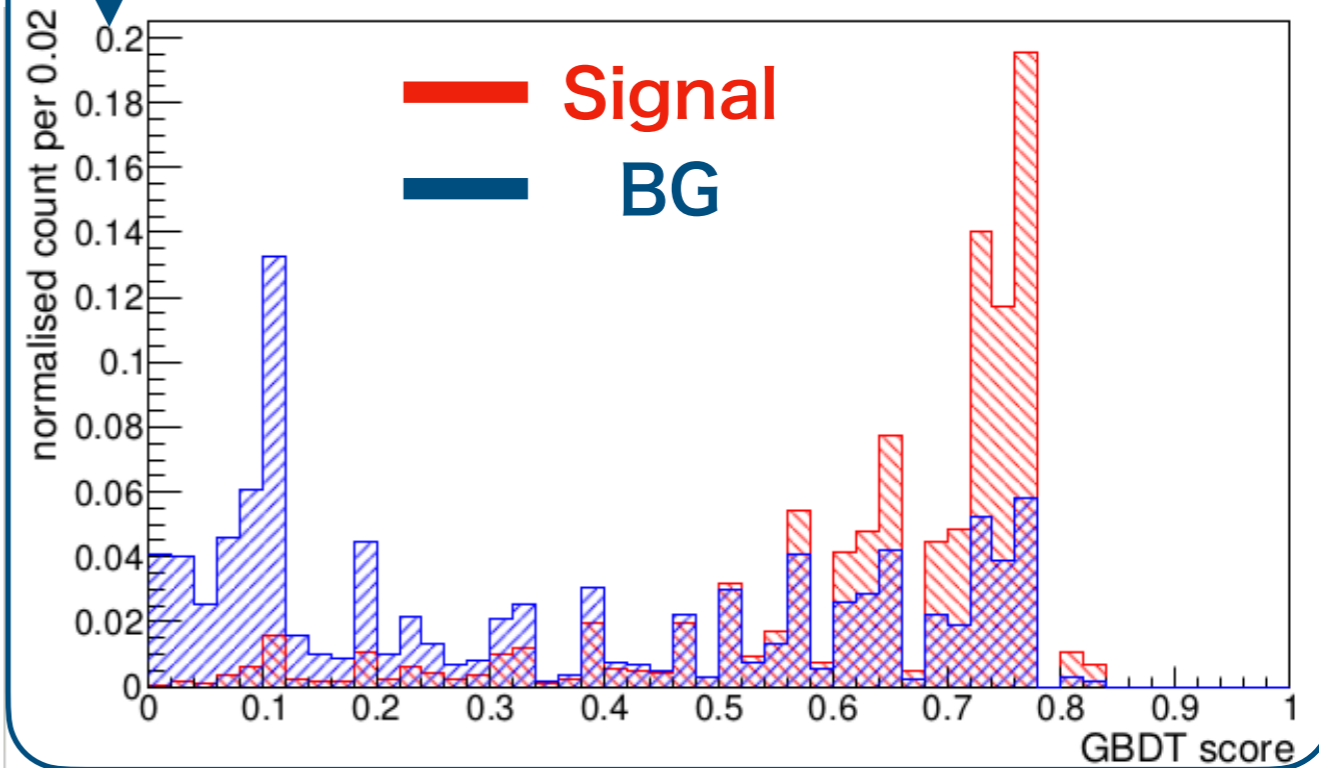
count

ADC sum compressed into 2bits

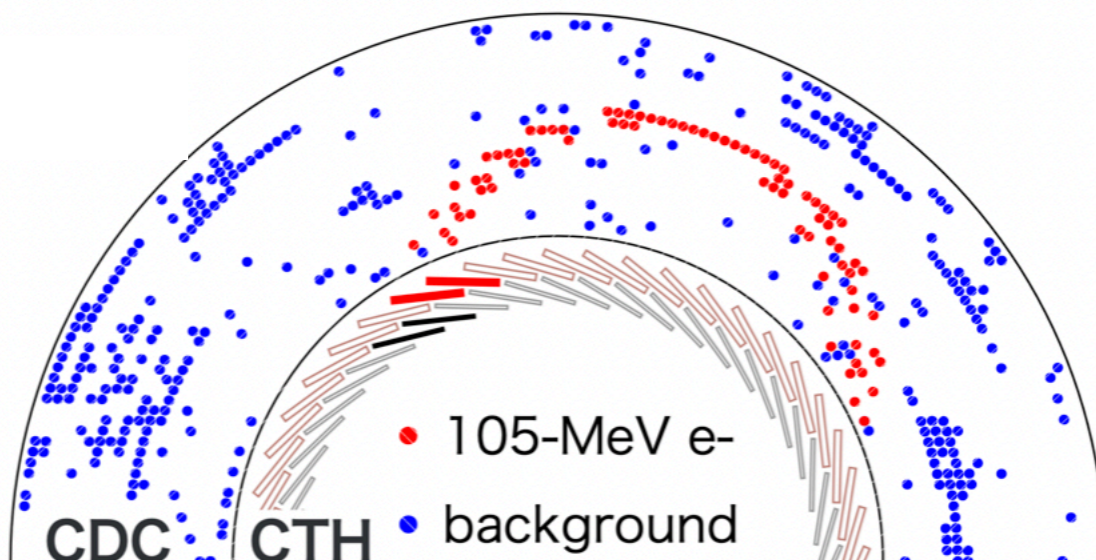
— signal

— Background

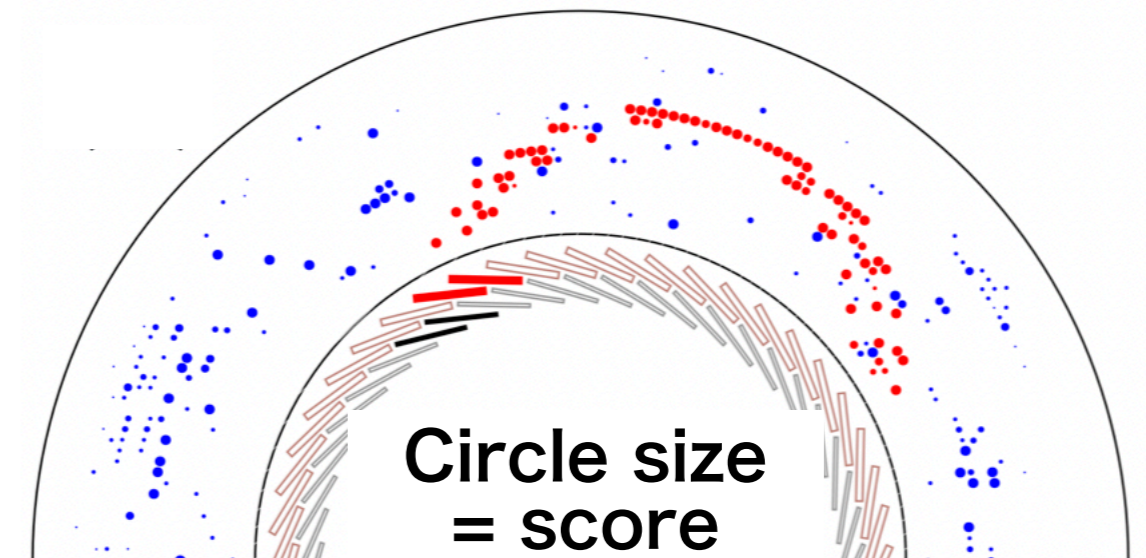
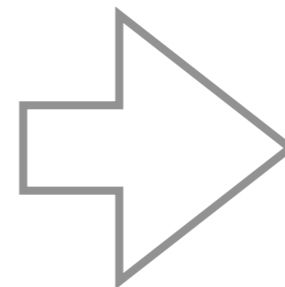
GBDT output score distribution



Hit map of signal event

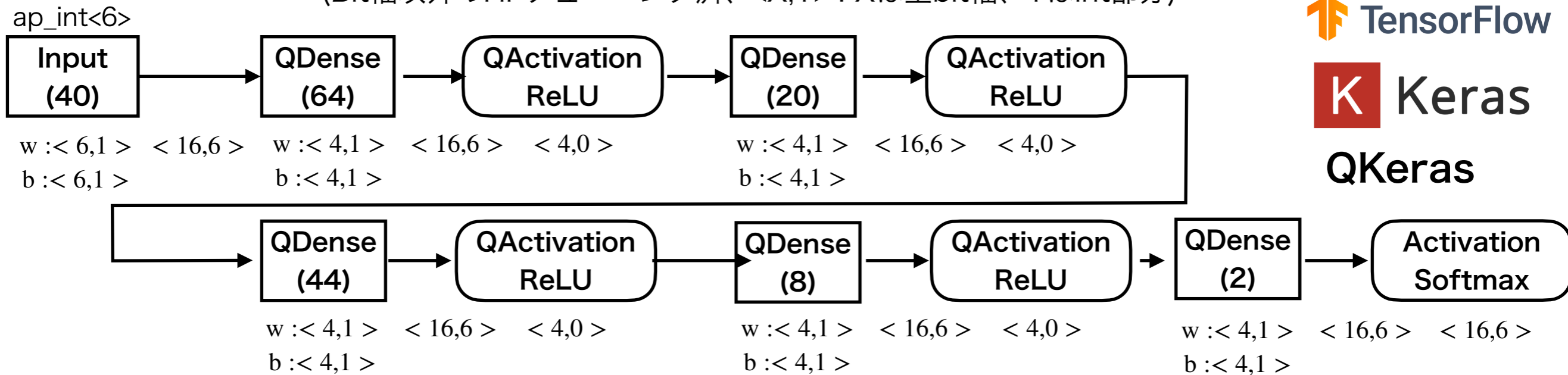


Scoring By GBDT



	CDC config	BG hit 占有率	ワイヤーヒット score情報	Active section
現実	20 layer x ~250 cell	~20 %	6 bit	~1500 ch
本スタディ	18 layer x 180 cell	5%	1 bit	960 ch

(Bit幅以外のHPチューニング済、 $\langle X, Y \rangle$: Xは全bit幅、Yはint部分)

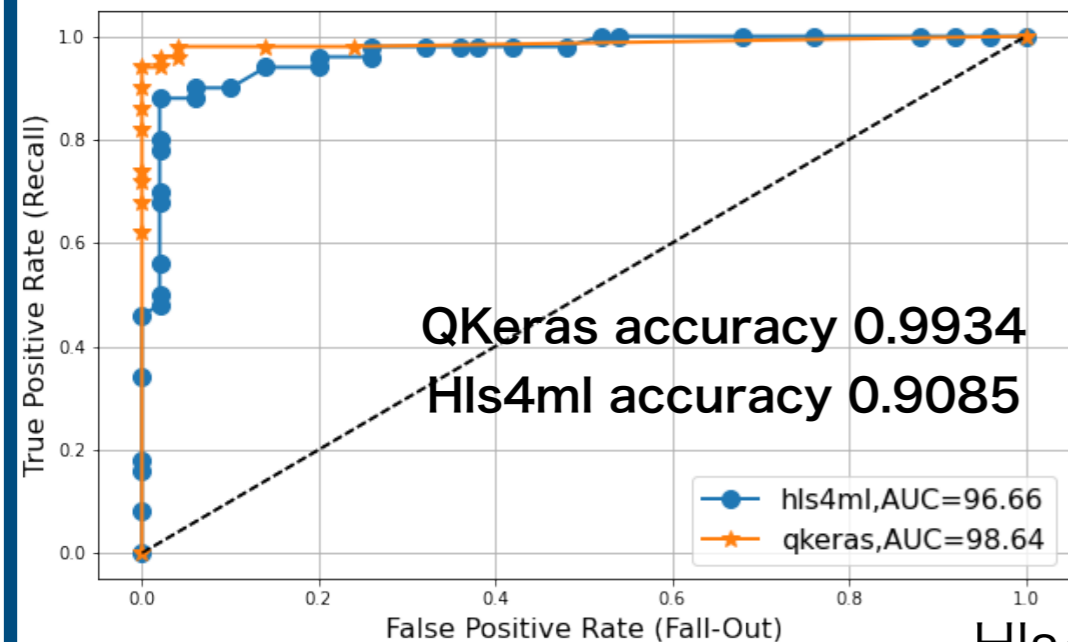


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モデルをハードウェア記述言語によるプログラミングなしに
ファームウェア変換可能なコード(RTL)に変換

<https://dx.doi.org/10.1088/1748-0221/13/07/P07027>

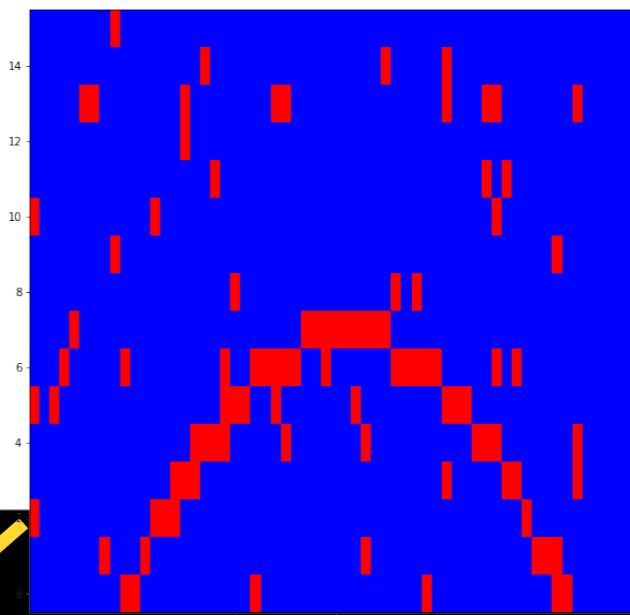
FPGA : AMD Xilinx Kintex-7 xck355t-ffg901-1



Latency @200MHz	Usage (%)			
	BRAM	DSP	FF	LUT
130 ns	~0	~0	5	32

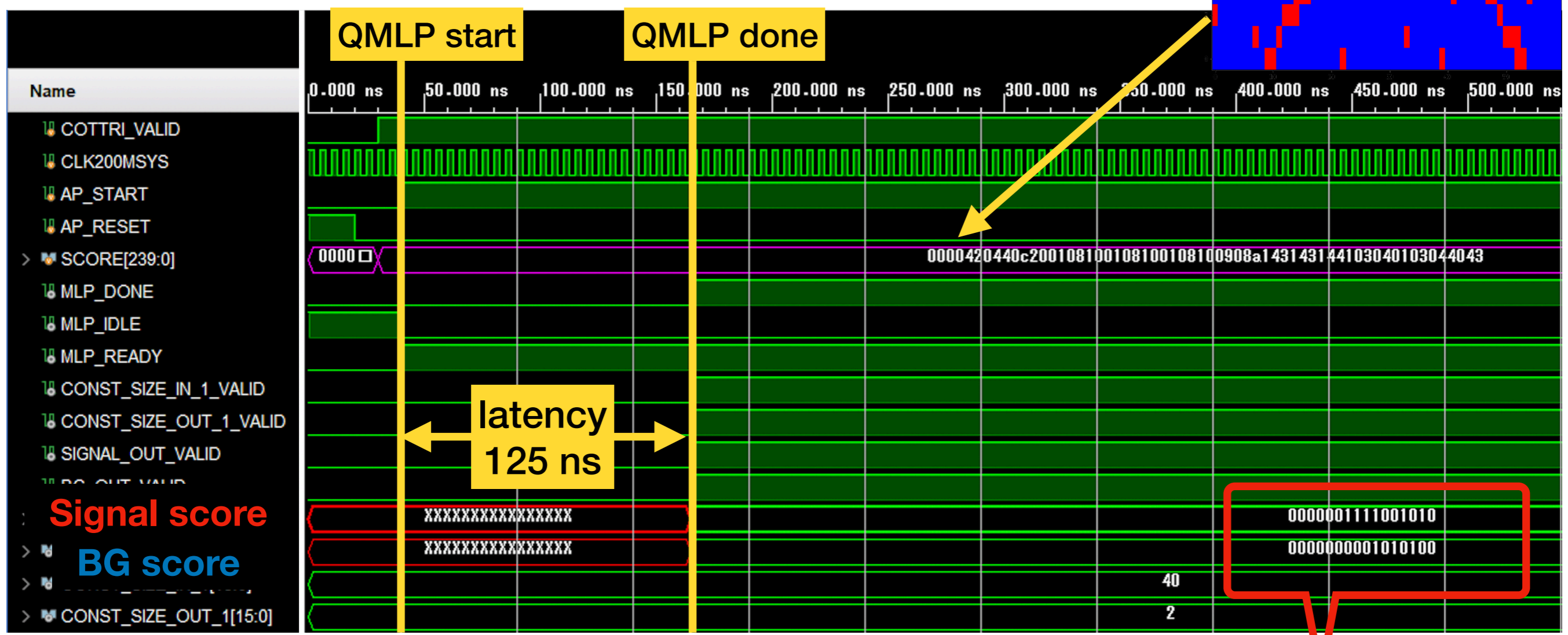
Hls4mlが生成したC++ファイルをVivado_hlsで高位合成した後、vivadoでこのQMLPのipを生成し、COTTRI MBのFWに組み込んだ

QMLP module simulation



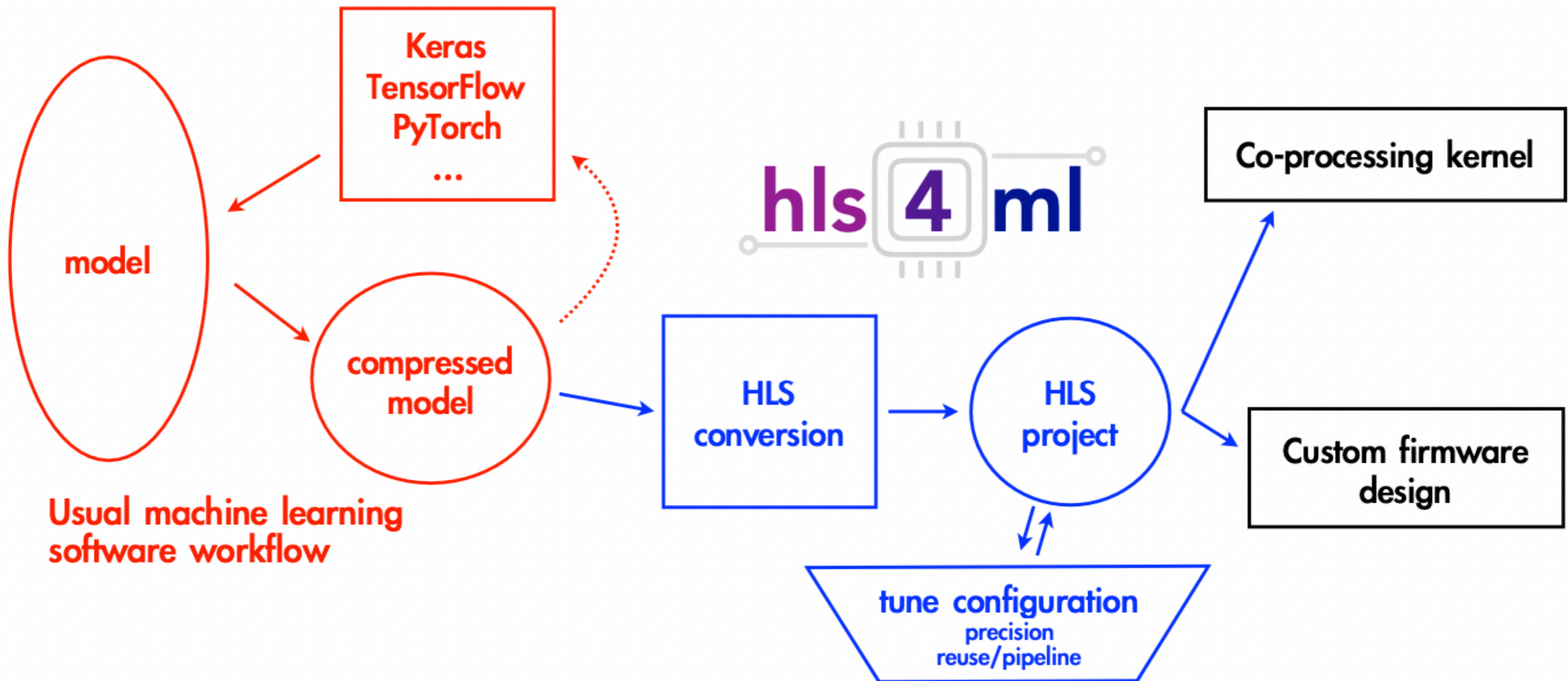
Xilinx vivado simulationによりQMLP moduleのlatencyと出力を確認

信号イベント情報を入力した時のwaveform



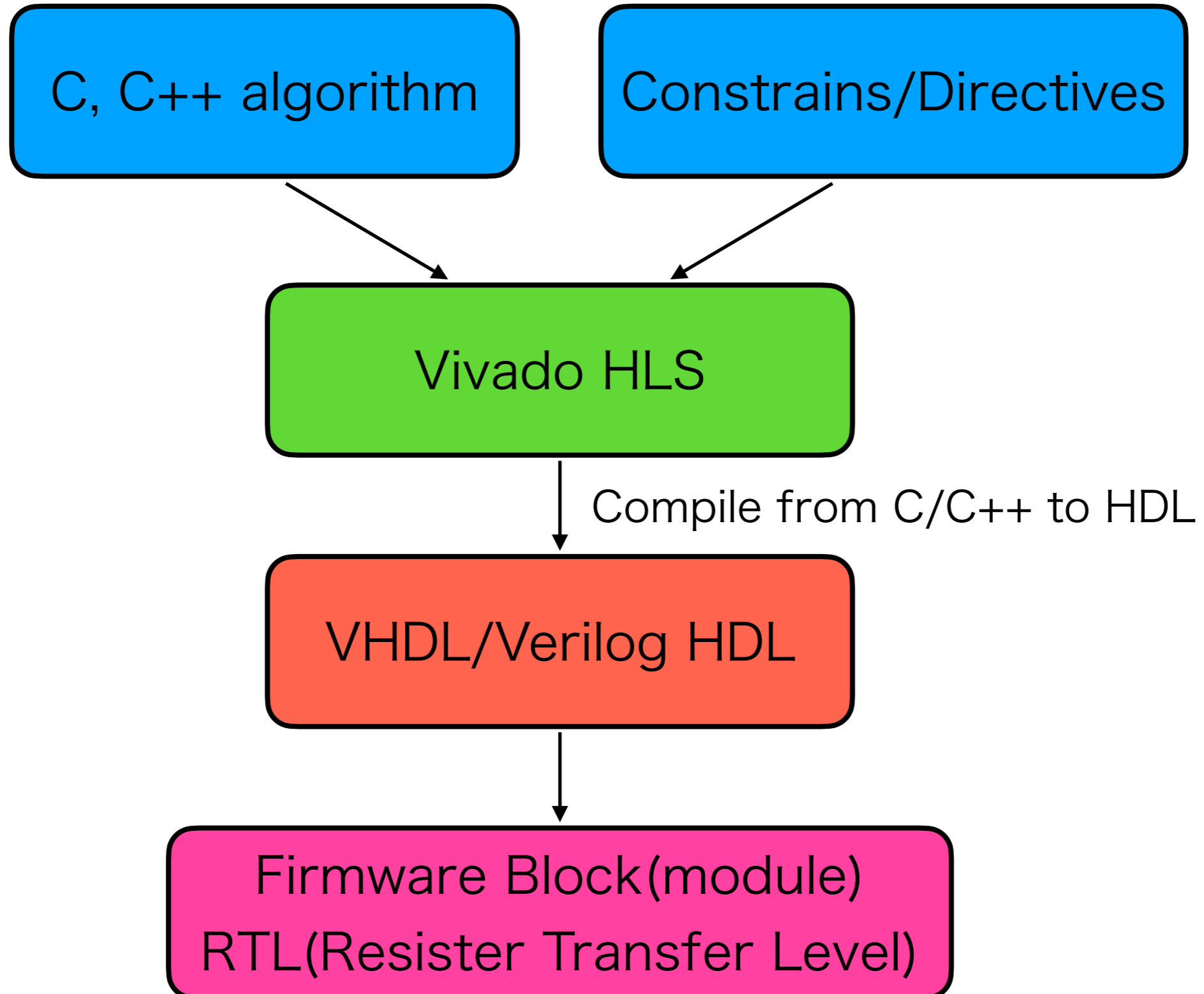
latency, score出力予想通り 😎

BG score < **Signal score**
0.08 0.95



Fast inference of deep neural networks in FPGAs for particle physics Fig1.

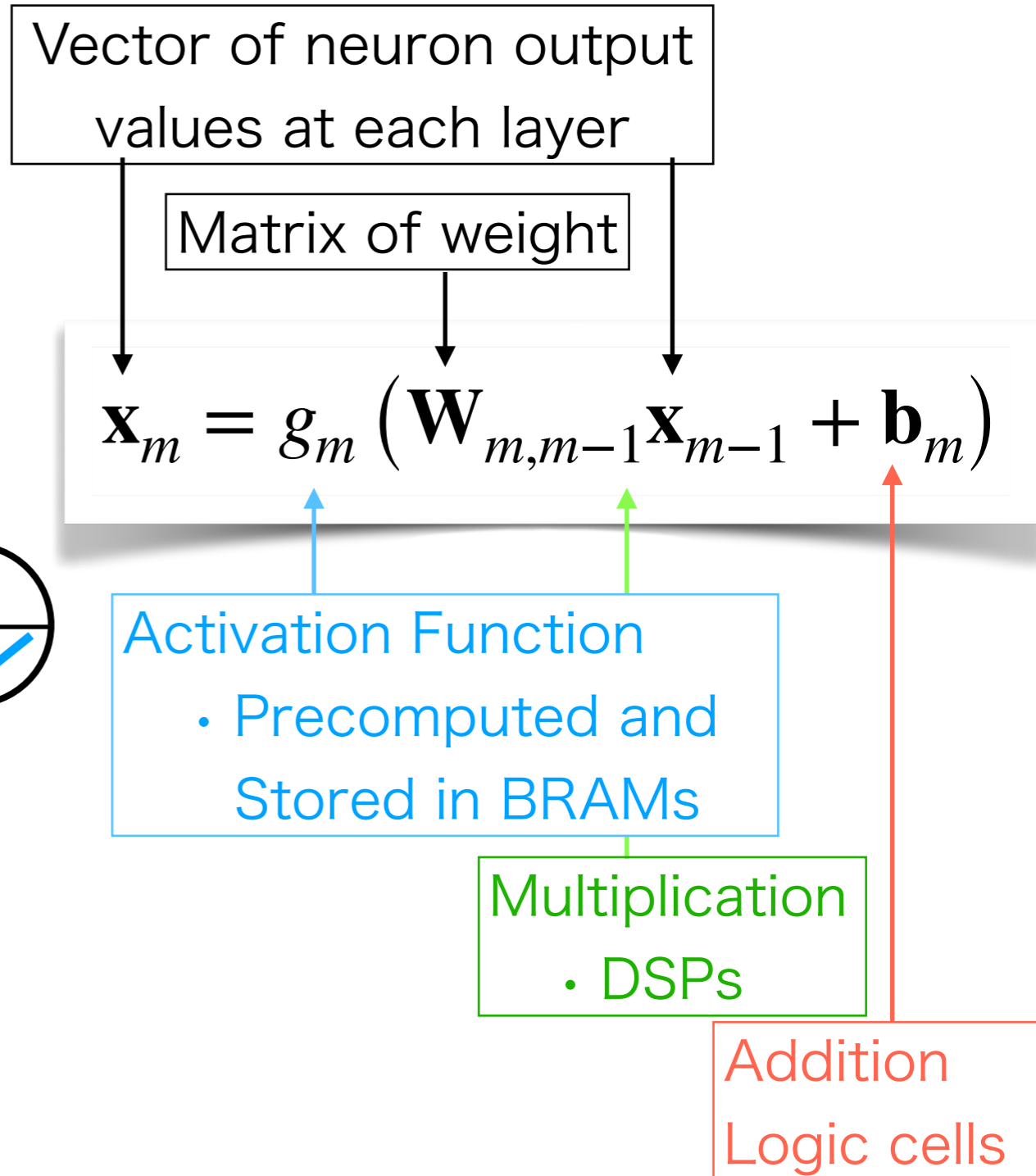
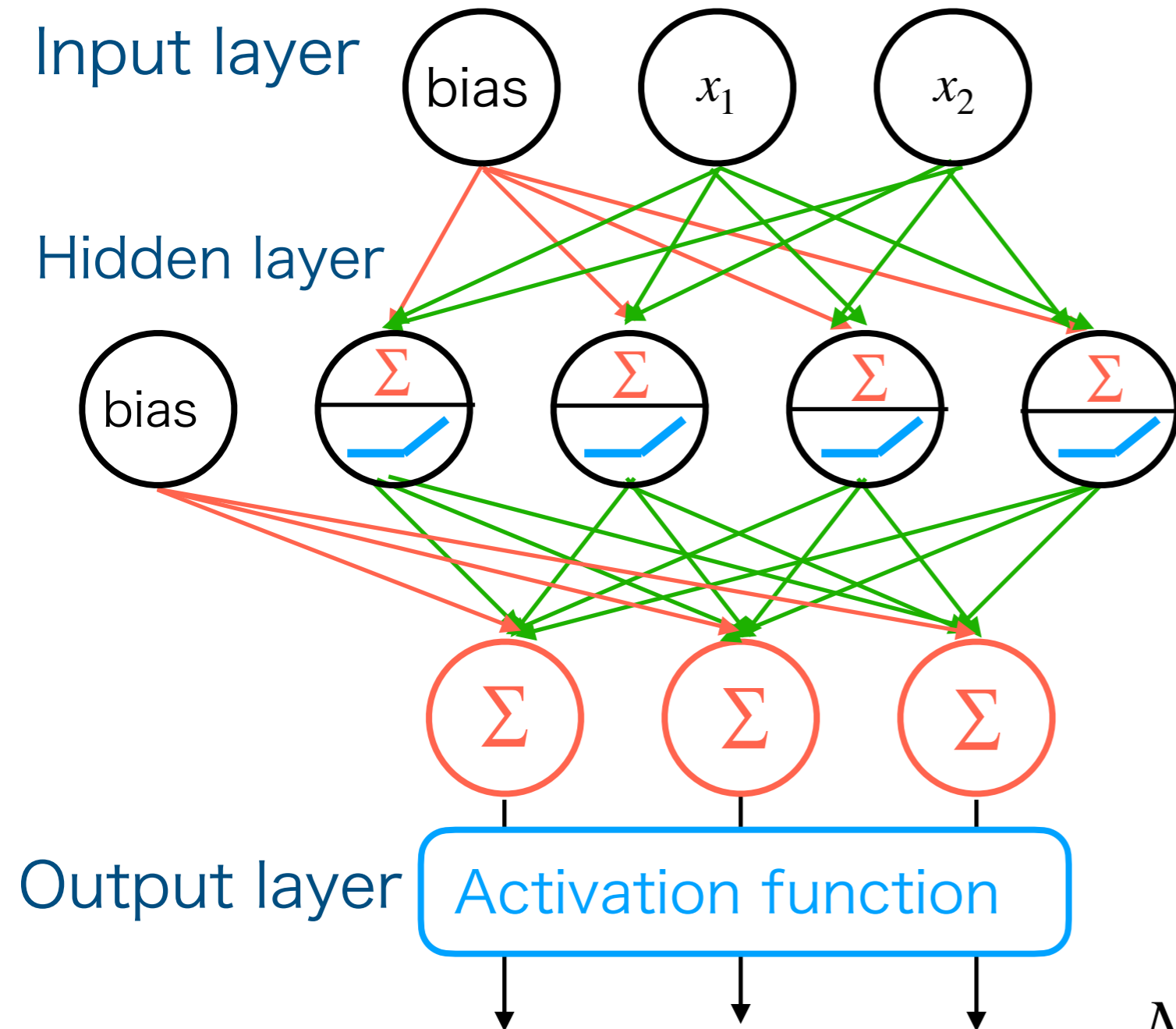
FPGA programming Flow



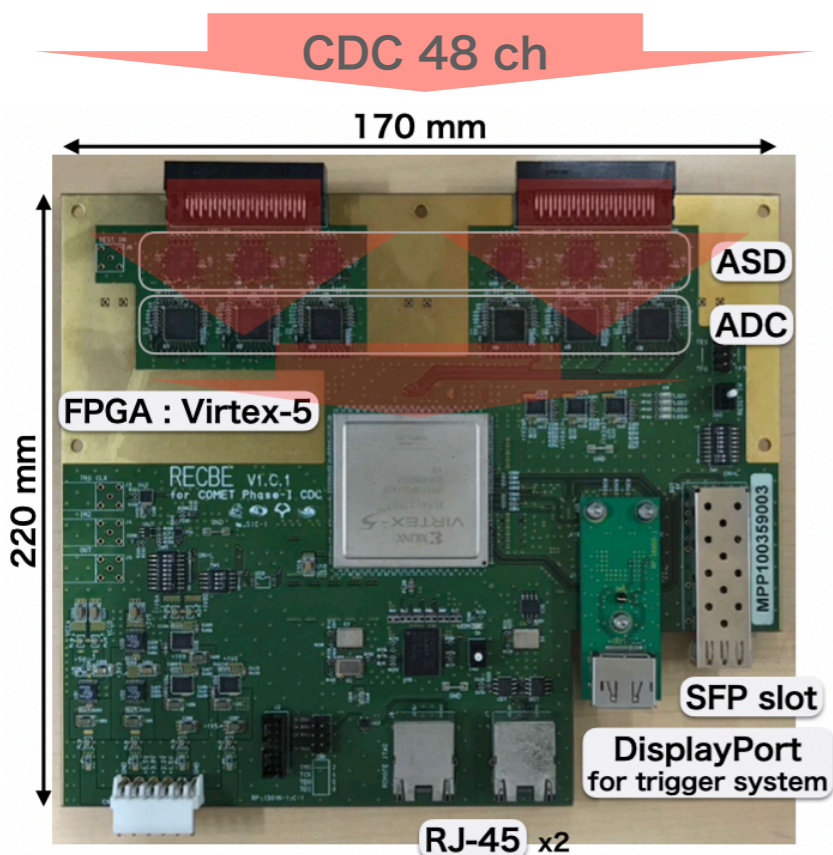
Neural Network on FPGA

Correspondence between Neural Network operation and FPGA resources

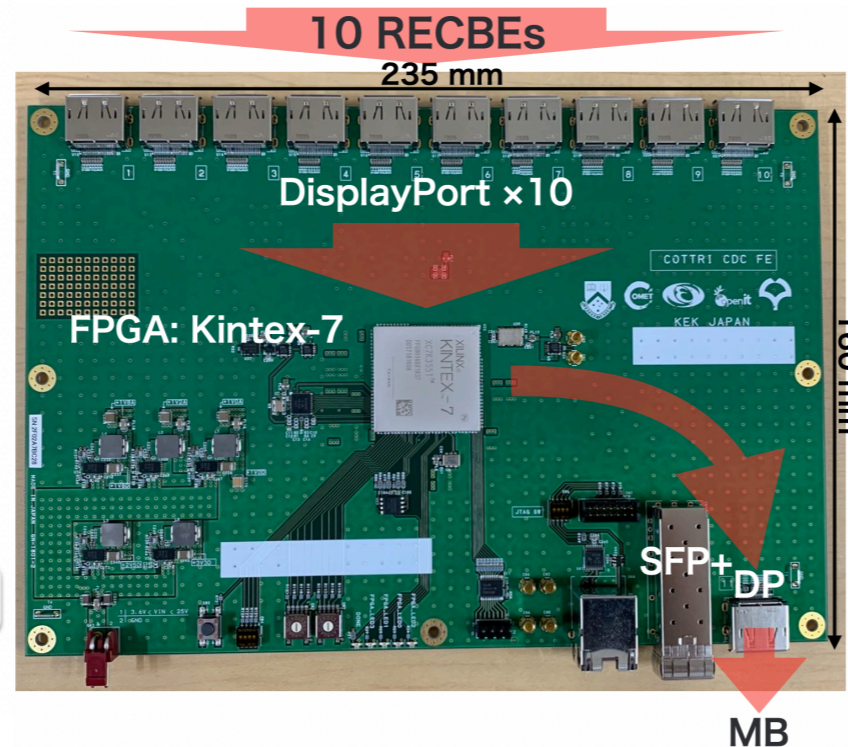
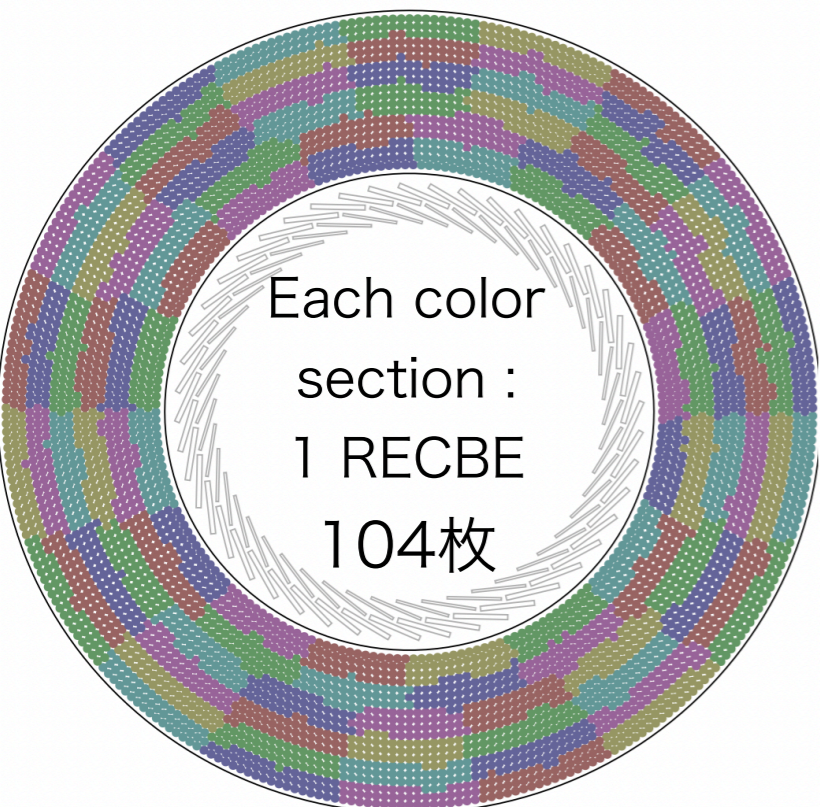
Ex) One-hidden-layer



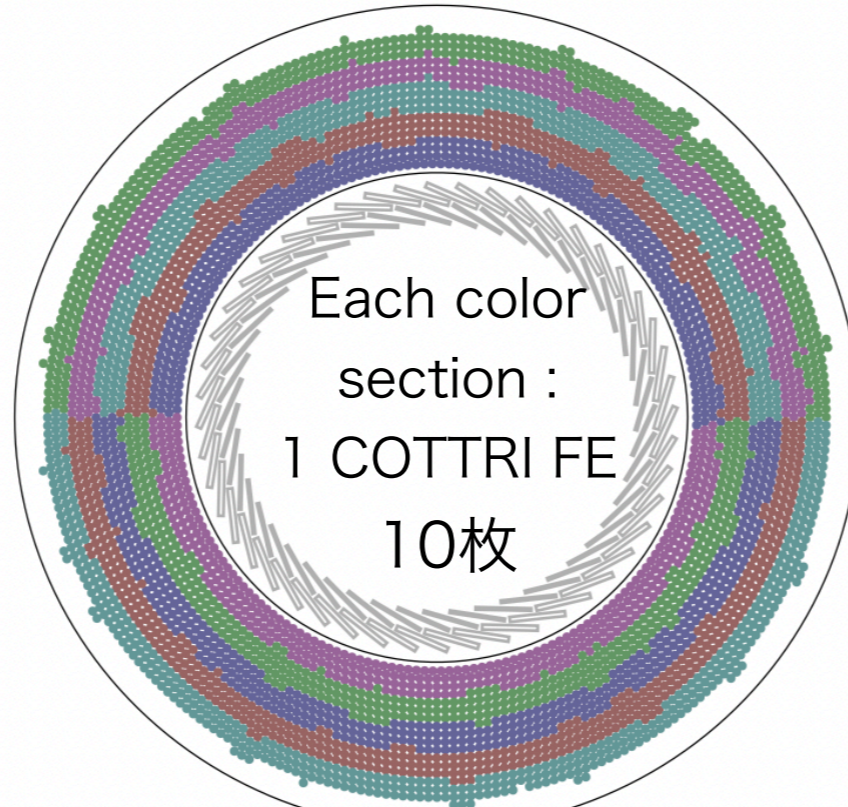
$$N_{multiplication} = \sum_{m=1}^M N_{m-1} \times N_m \propto \text{DSPs}$$



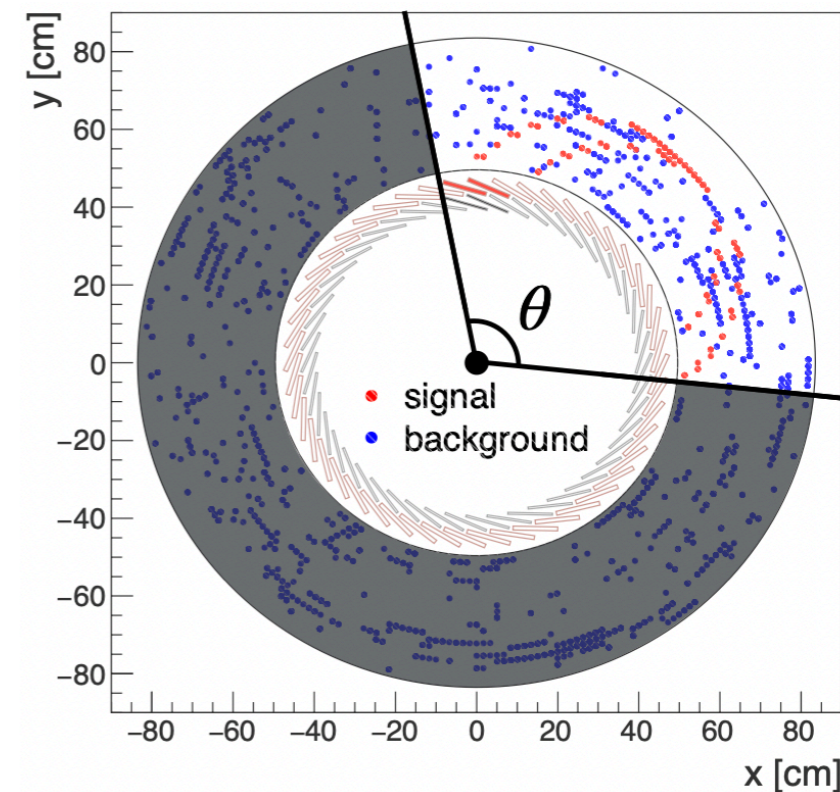
RECBE Configuration



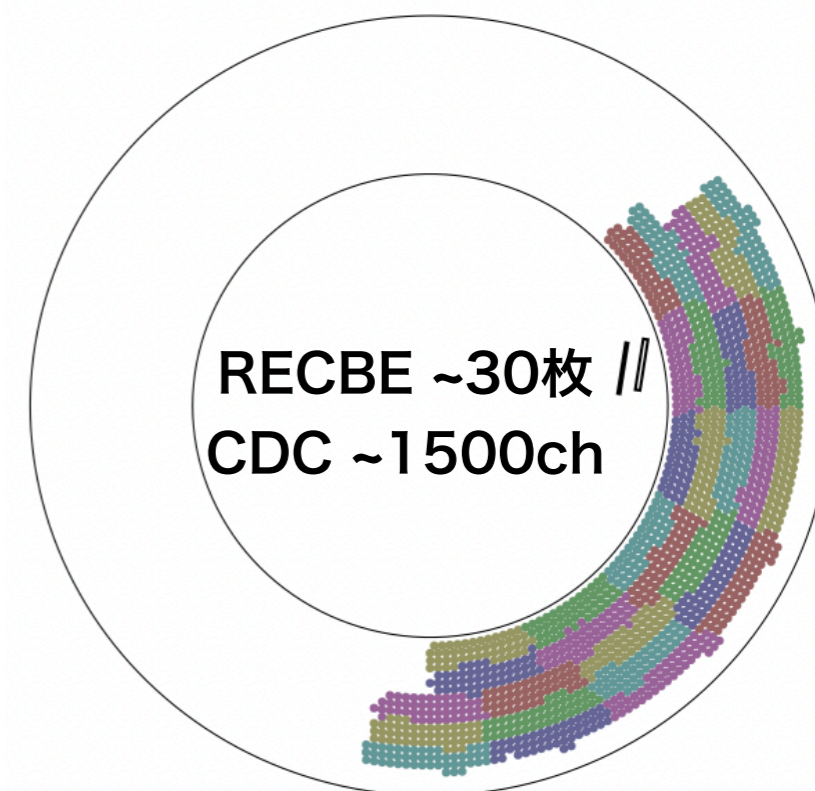
COTTRI FE Configuration



信号電子が軌跡を残すのは
CDCの約1/3領域



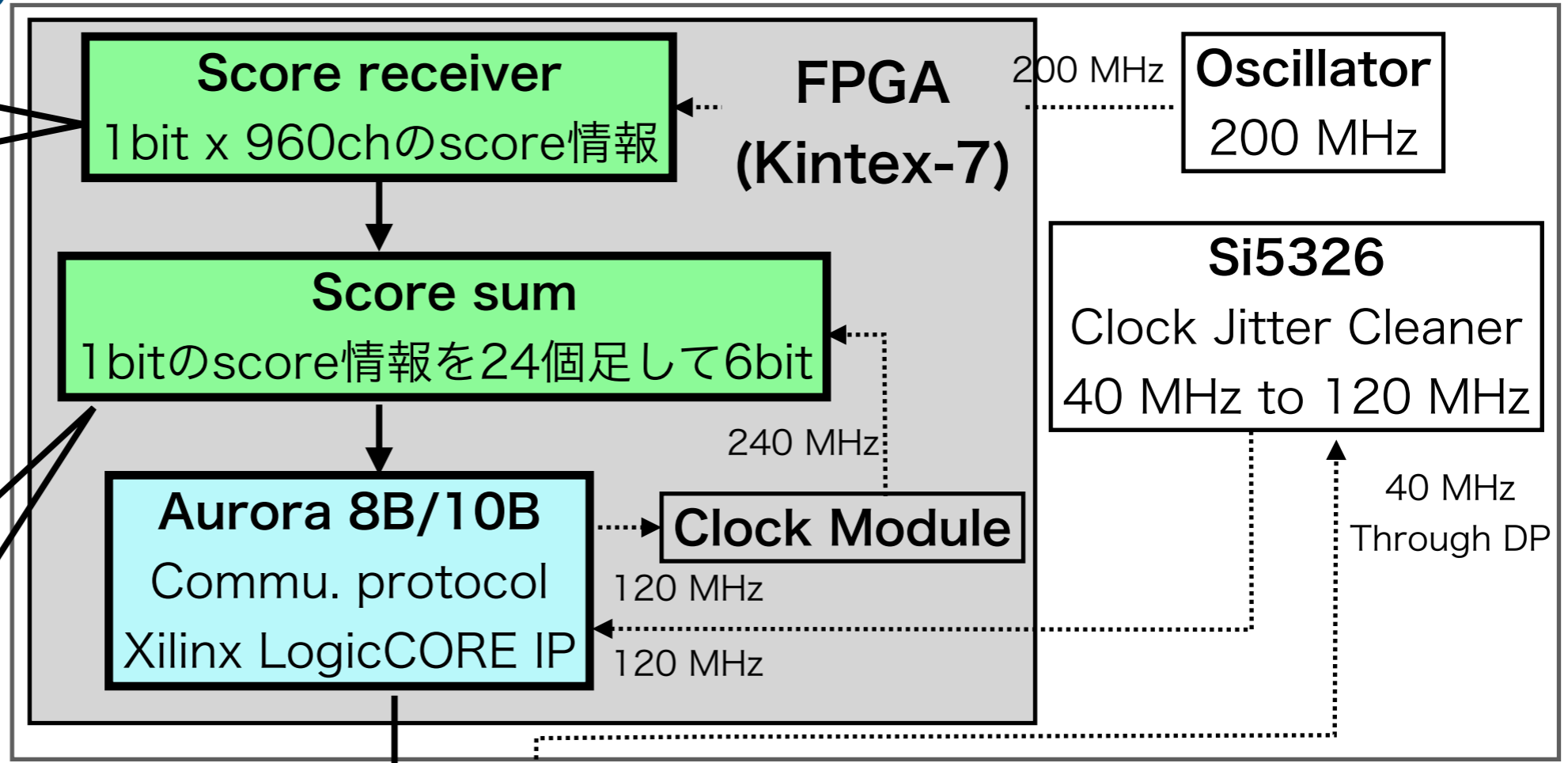
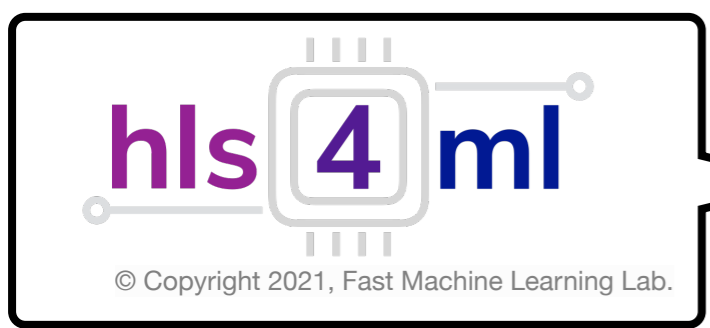
Active sectionの例



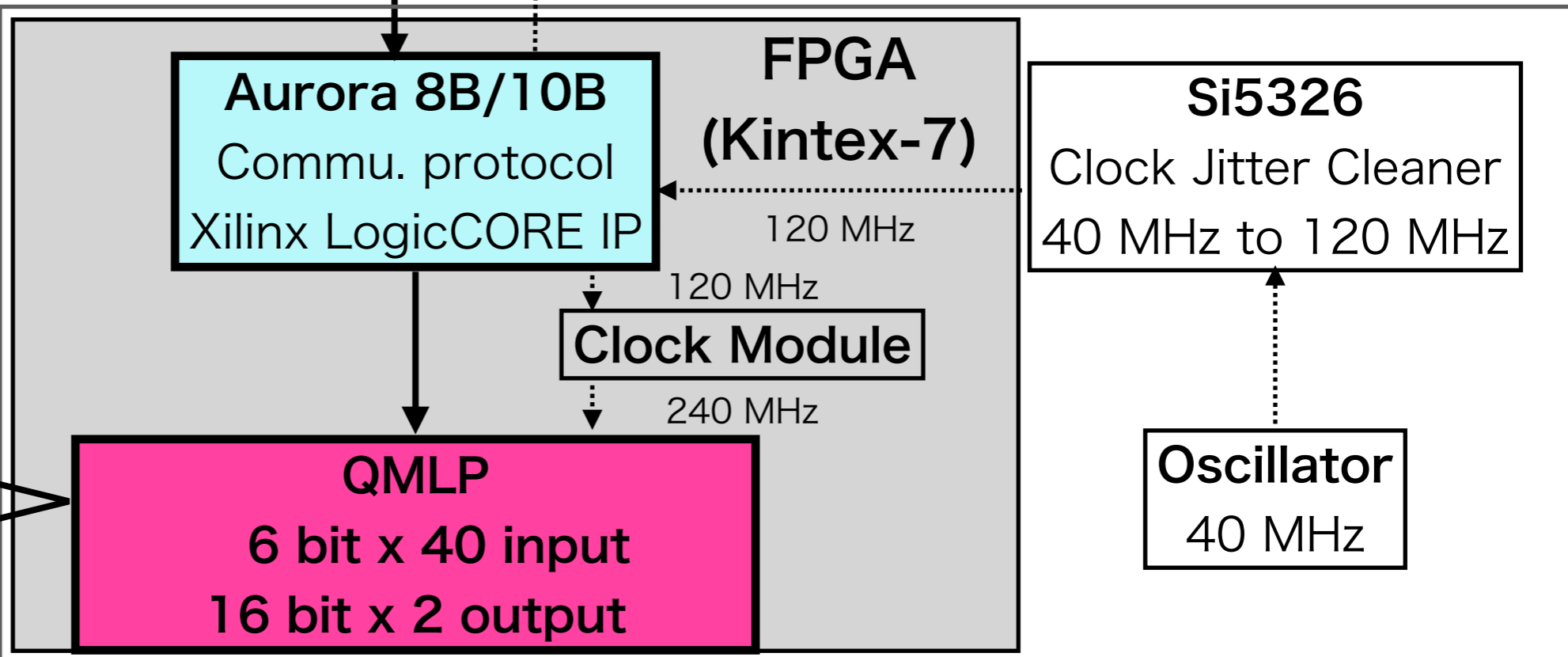
Firmware

Active section (CDC wire 960 ch) の1bit score情報をSiTCPでアドレスに書き込み

FE-MB間のデータ転送は10MHz。
960 bit のscoreを10MHzで転送できないので、6bit x 40 に圧縮



COTTRI Merger-Board



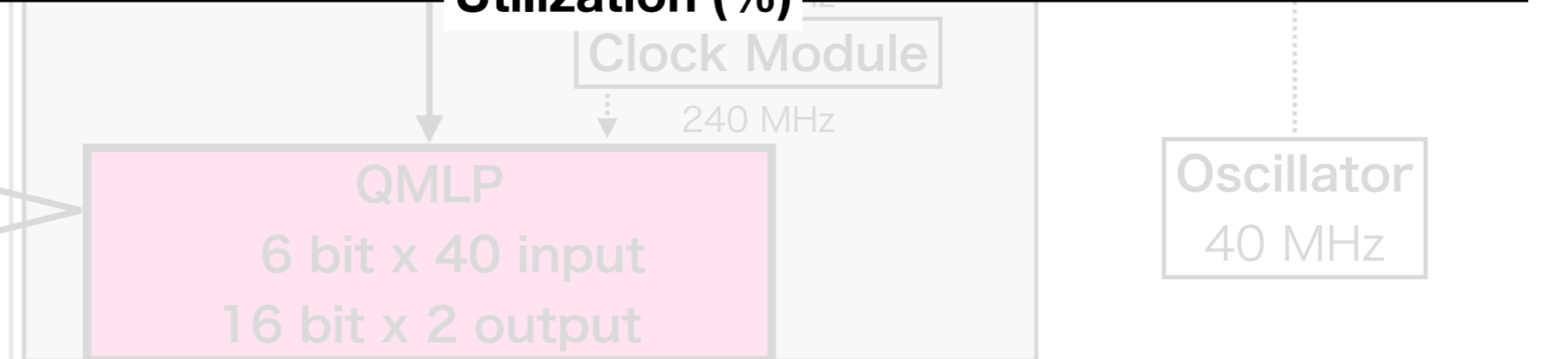
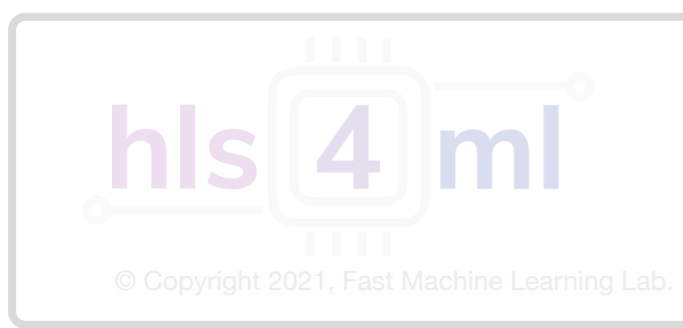
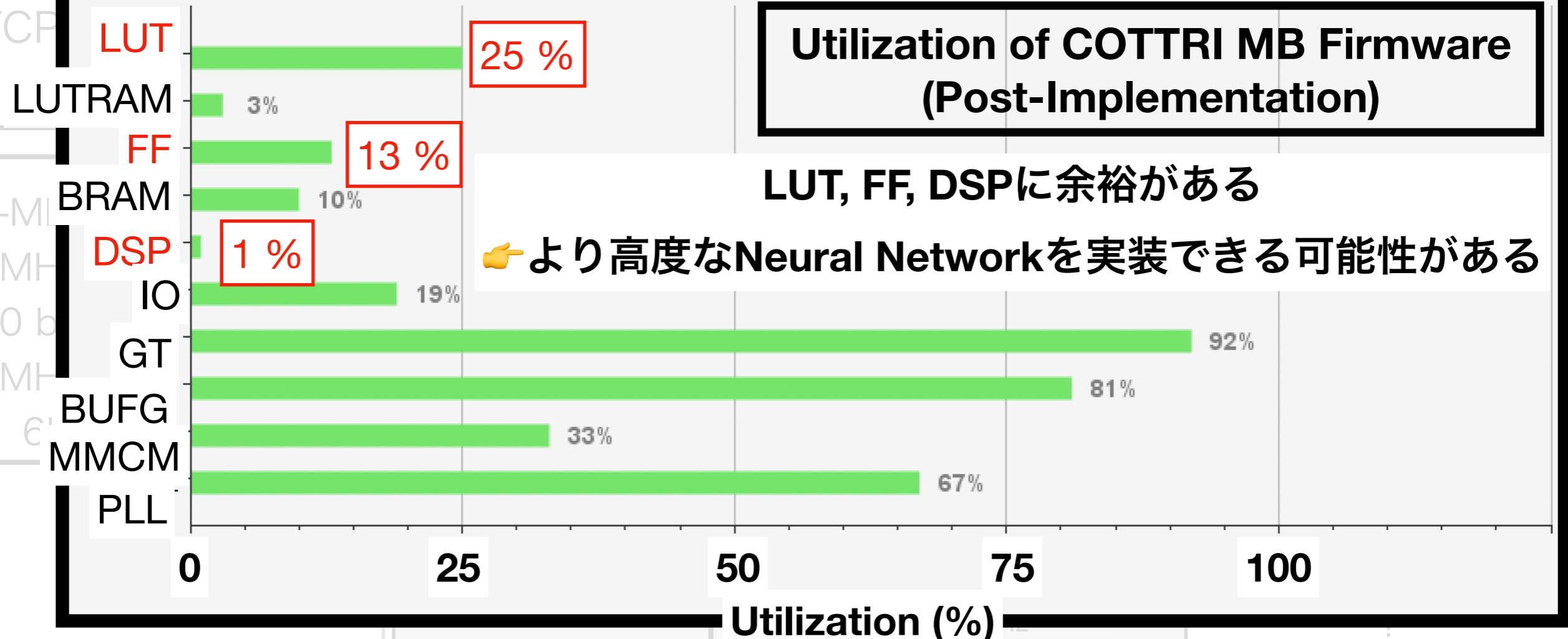
Active section
(CDC wire 960 ch)
の1bit score情報を
SiTCP



Utilization of COTTRI MB Firmware (Post-Implementation)

LUT, FF, DSPに余裕がある

👉 より高度なNeural Networkを実装できる可能性がある



- Maximum data transfer = 2.4 Gbps/lane x 2lane x 0.8 = 3.84 Gbps
- Data format : 1 header & 10 data packets

1 frame @ 10 MHz

Bit	31	30	29	28	27	26	25	24	23	22	21	20	19	18	17	16	15	14	13	12	11	10	9	8	7	6	5	4	3	2	1	0
Header	0	Parity bits			1	Sent number												Board ID														
Score	0	Parity bits			0	RECBE 9																										
	0	Parity bits			0	RECBE 8																										
	0	Parity bits			0	RECBE 7																										
	0	Parity bits			0	RECBE 6																										
	0	Parity bits			0	RECBE 5																										
	0	Parity bits			0	RECBE 4																										
	0	Parity bits			0	RECBE 3																										
	0	Parity bits			0	RECBE 2																										
	0	Parity bits			0	RECBE 1																										
	0	Parity bits			0	RECBE 0																										

COTTRI FE to COTTRI MB data format

For the preliminary study

1 frame @ 10 MHz

Bit	31	30	29	28	27	26	25	24	23	22	21	20	19	18	17	16	15	14	13	12	11	10	9	8	7	6	5	4	3	2	1	0
Header	0	Parity bits			1	Sent number											Board ID															
Score	0	Parity bits			0	Input39			Input38			Input37			Input36																	
	0	Parity bits			0	Input35			Input34			Input33			Input32																	
	0	Parity bits			0	Input31			Input30			Input29			Input28																	
	0	Parity bits			0	Input27			Input26			Input25			Input24																	
	0	Parity bits			0	Input23			Input22			Input21			Input20																	
	0	Parity bits			0	Input19			Input18			Input17			Input16																	
	0	Parity bits			0	Input15			Input14			Input13			Input12																	
	0	Parity bits			0	Input11			Input10			Input9			Input8																	
	0	Parity bits			0	Input7			Input6			Input5			Input4																	
	0	Parity bits			0	Input3			Input2			Input1			Input 0																	

$$precision = \frac{TP}{TP + FP}$$

陽性予測の正解率

$$recall = \frac{TP}{TP + FN}$$

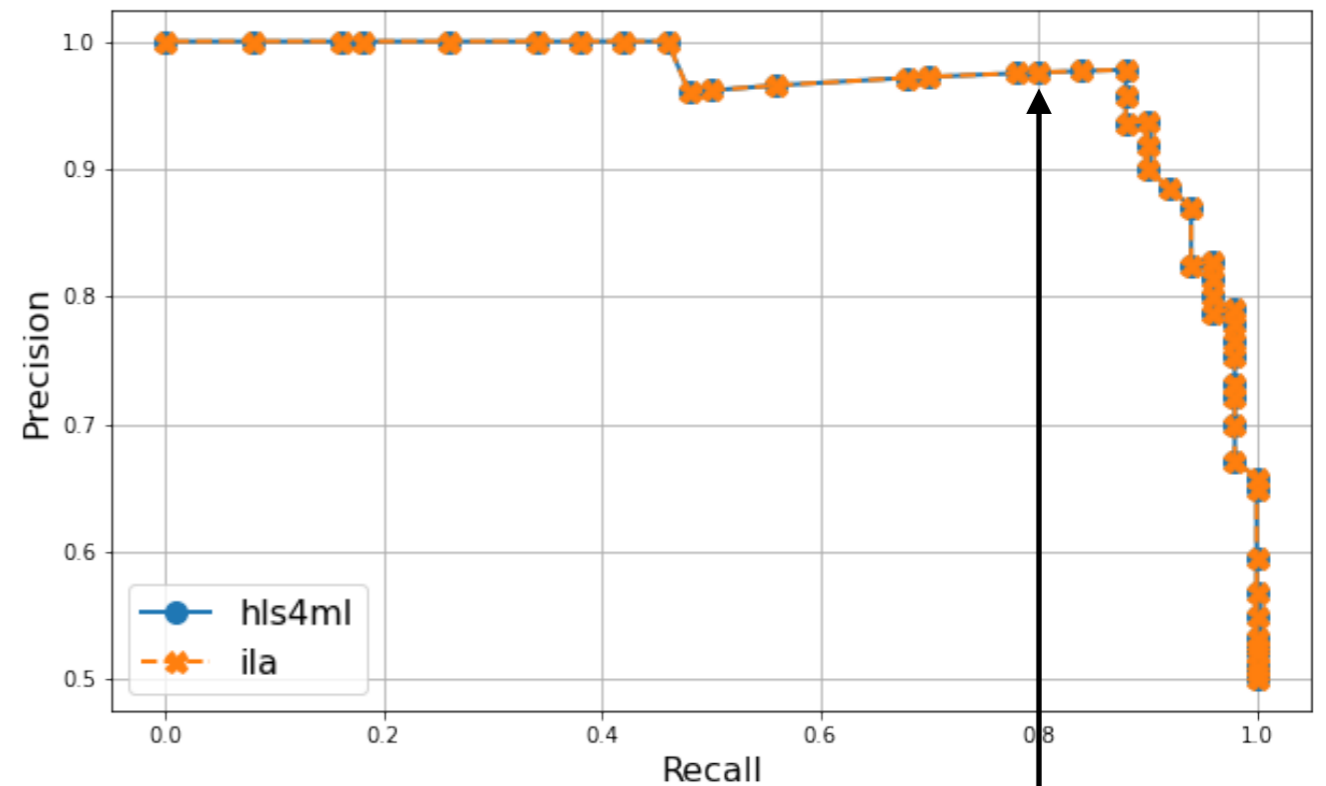
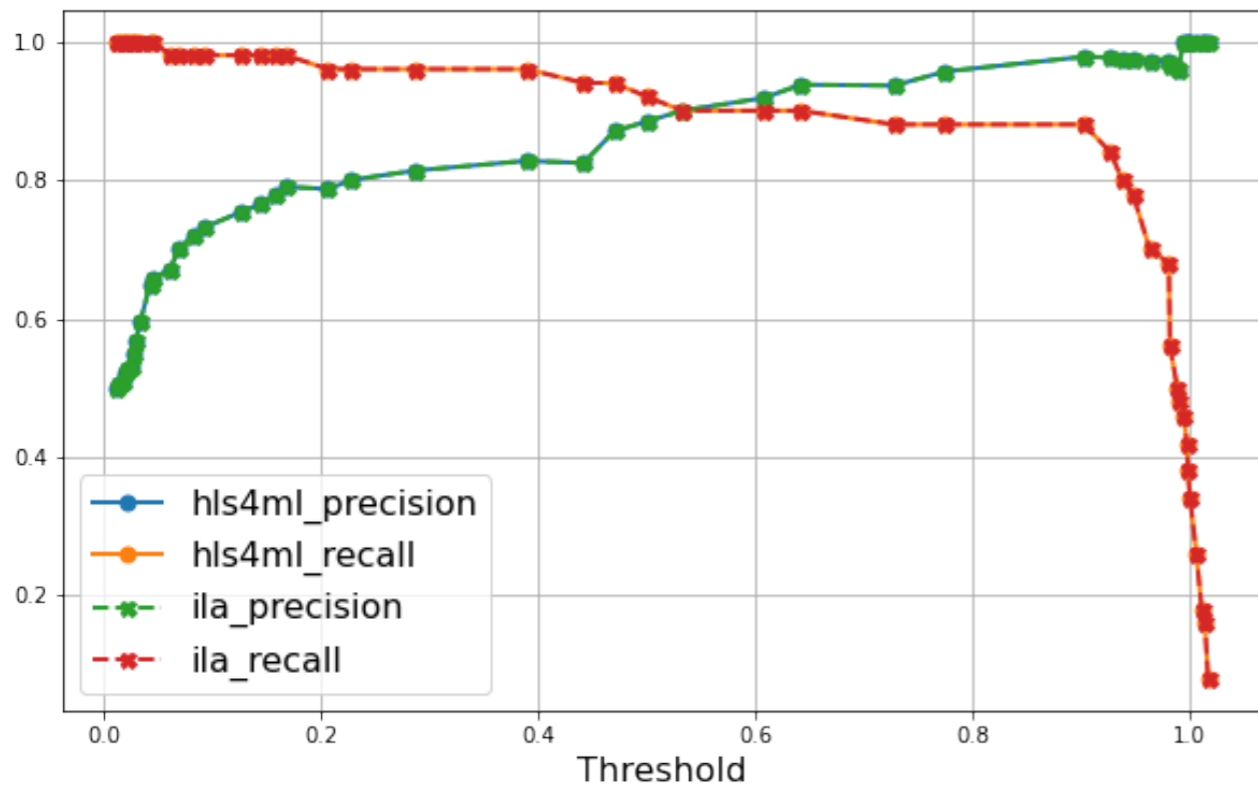
正しく分類した陽性の割合

TP : 真陽性の数 Signal eventをsignal eventであると分類した数

TN : 新陰性の数 BG eventをBG eventであると分類した数

FP : 偽陽性の数 BG eventをsignal eventであると分類した数

FN : 偽陰性の数 Signal eventをBG eventであると分類した数



分類器がSignal eventであると予測したとき、その予測が正しいのは98%
 全てのSignal eventの80%を検出

$$TPR = \frac{TP}{TP + FN}$$

正しく分類した陽性の割合

$$FPR = \frac{FP}{TN + FP}$$

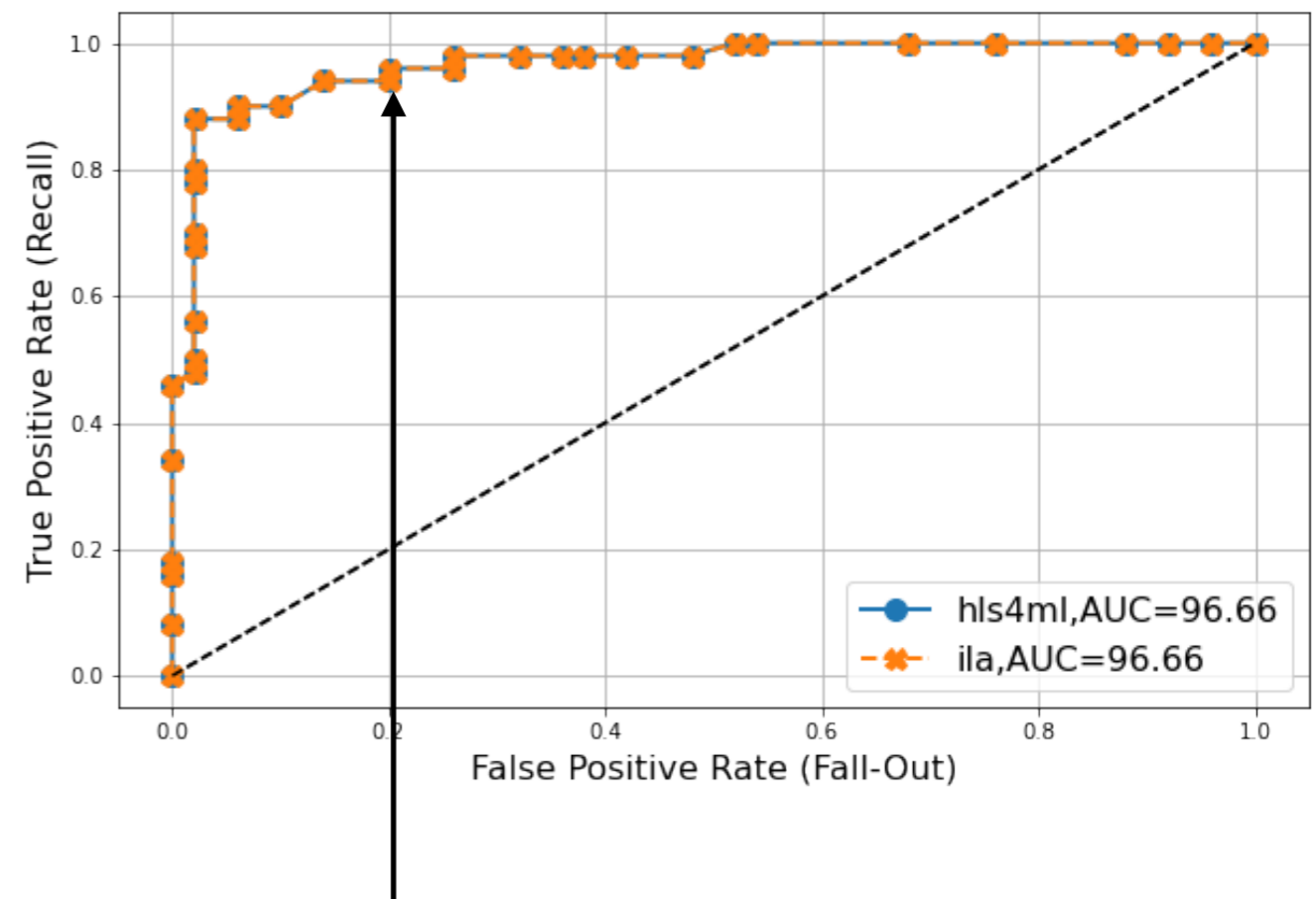
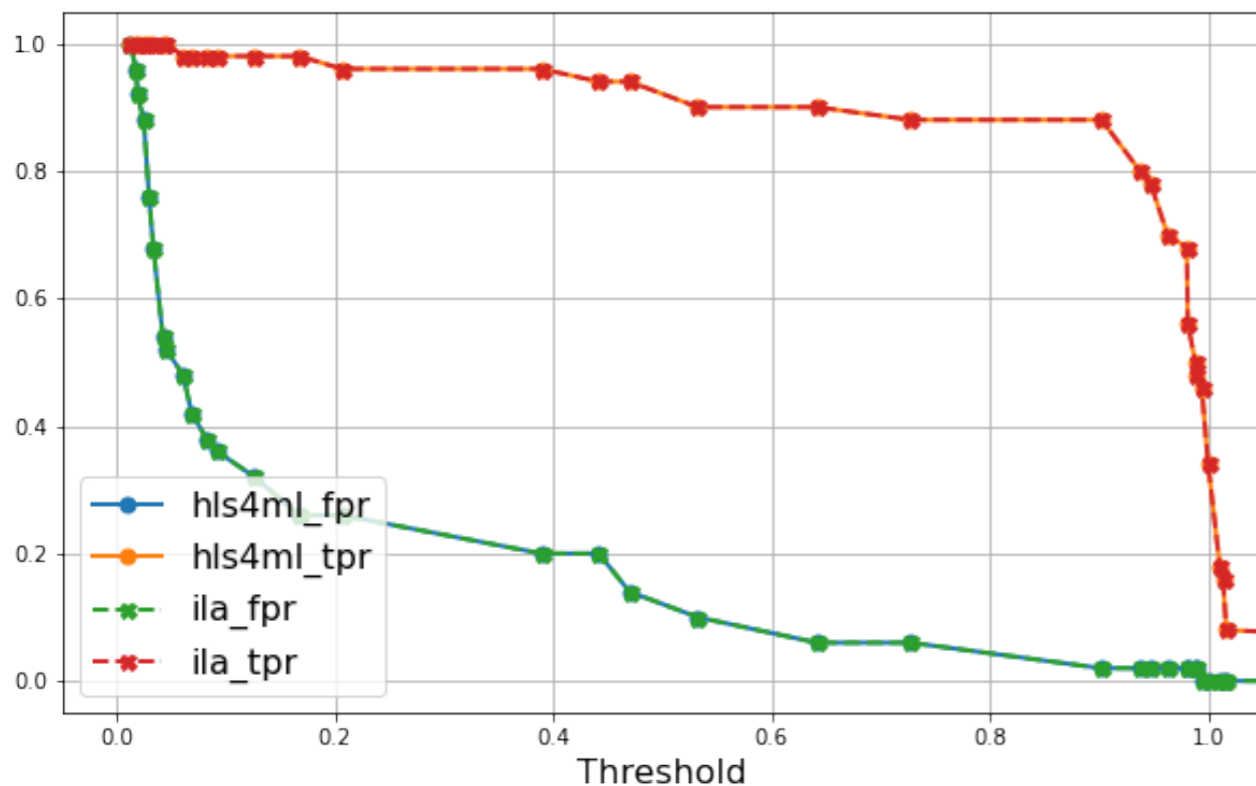
誤って陽性と分類された陰性の割合

TP : 真陽性の数 Signal eventをsignal eventであると分類した数

TN : 新陰性の数 BG eventをBG eventであると分類した数

FP : 偽陽性の数 BG eventをsignal eventであると分類した数

FN : 偽陰性の数 Signal eventをBG eventであると分類した数



分類器がSignal eventであると予測したとき、その予測が正しいのは98%

全てのBG eventの20%を誤ってSignal eventであると分類