Contributions to the ePIC DAQ Project

TAKU GUNJI CENTER FOR NUCLEAR STUDY THE UNIVERSITY OF TOKYO



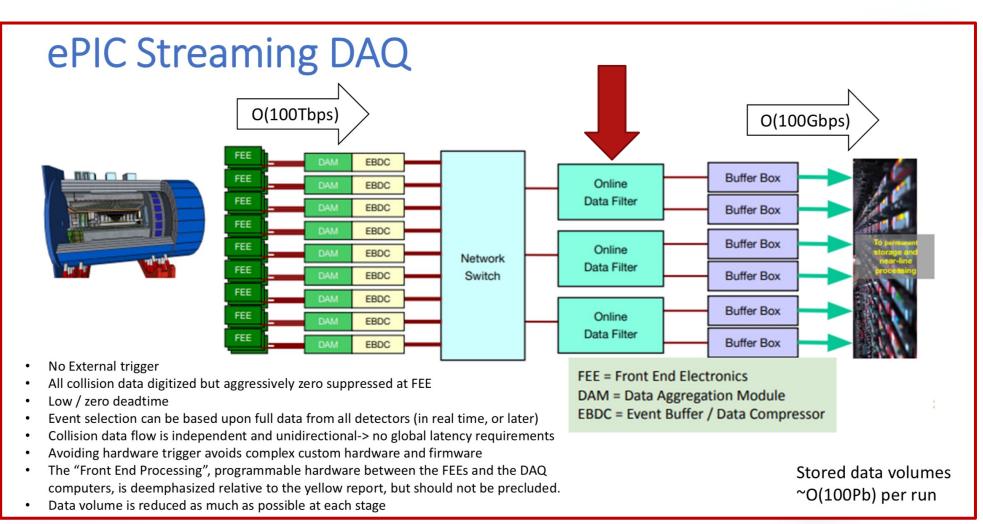


<u>Outline</u>

Free Streaming Readout
ALICE O2
Some (small) activities
SPADI-Alliance
Summary

Free Streaming Readout

Will be a future standard DAQ system

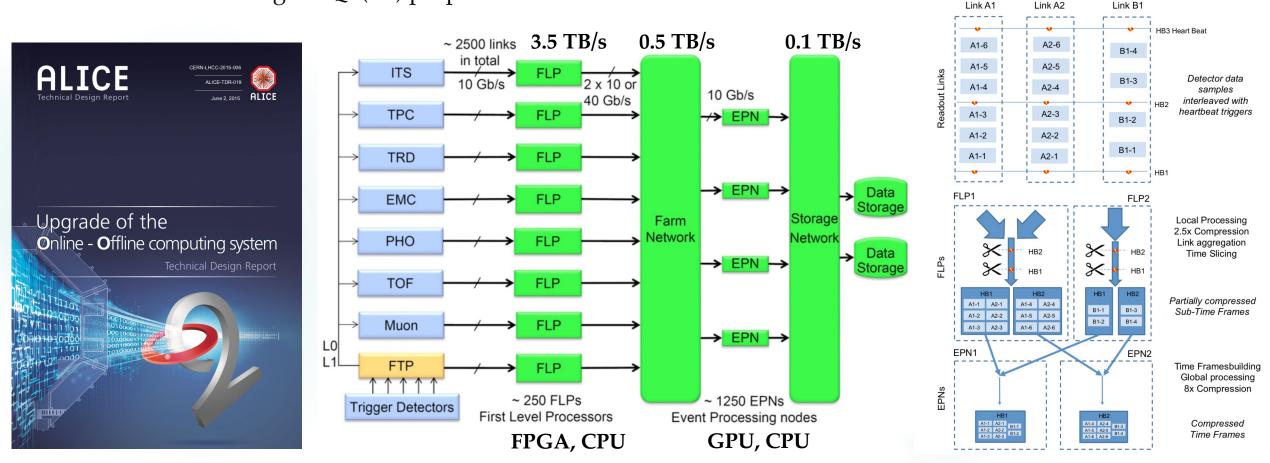


ALICE O² system

50kHz Pb-Pb collisions

TPC moved from triggered readout (gating grid) to continuous readout with GEMs ALICE had experiences on hardware accelaration (GPUs) in HLT during Run2

 \rightarrow New Free streaming DAQ (O²) proposed in 2015



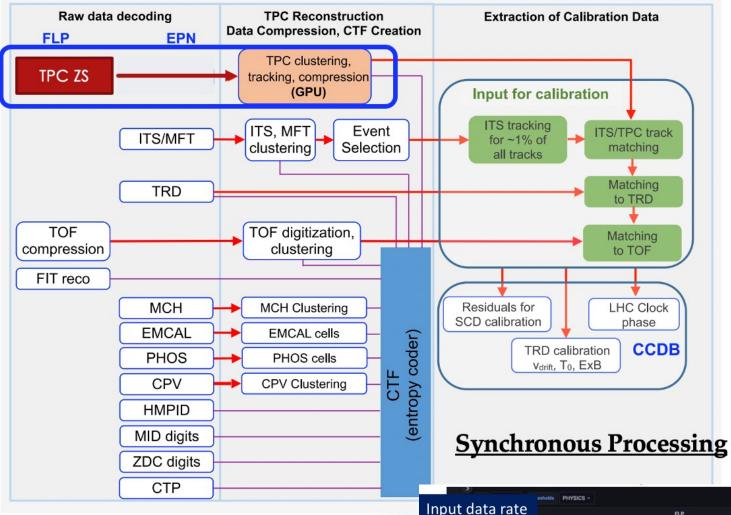
Detector B

Detector A

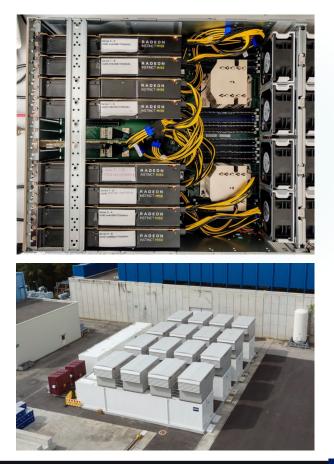
Link A2

Link A1

Online processing in ALICE O²



8 GPU/EPN, 250 EPNs



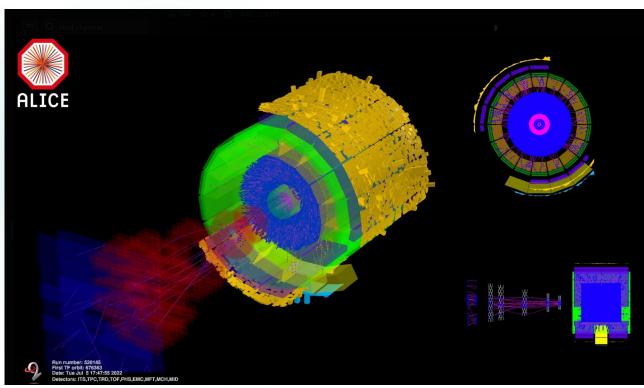
Compressed Raw Data Data FLP DPL In DPL Out StfSender Out TfBuilder In TfBuilder Out DPL In CTF writer StfBuilder StfSender In 95.6 GB/s 95.0 GB/s 93.1 GB/s 91.9 GB/s 94.0 GB/s 109 GB/s 70.9 GB/s 85.4 GB/s 93.5 GB/s 868 MB/s 90.4 GB/s

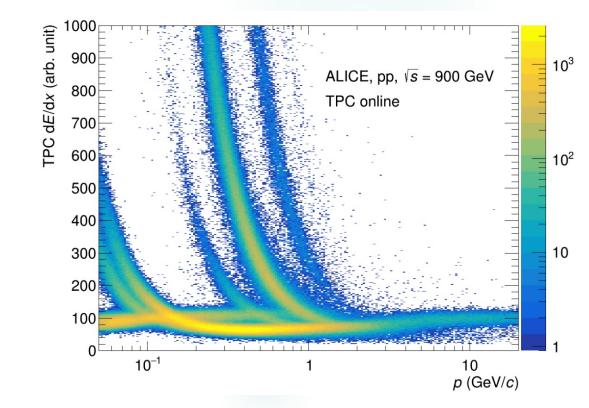
Online processing performance

Online clustering, tracking, PID

All are done ONLINE

Event display after online reconstruction 1 TF = 11.5 ms, 10 kHz IR \rightarrow ~100 collisions

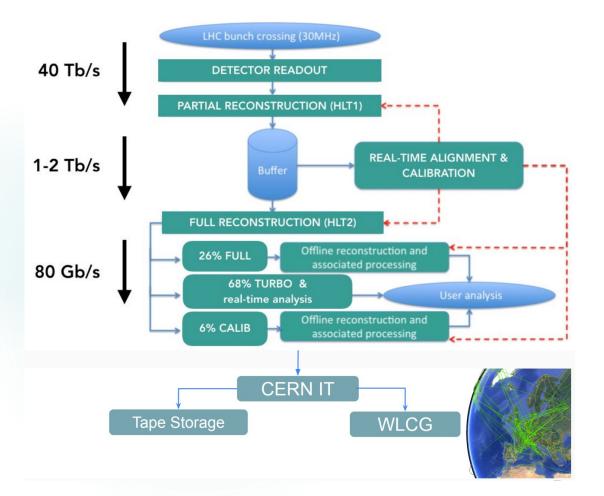


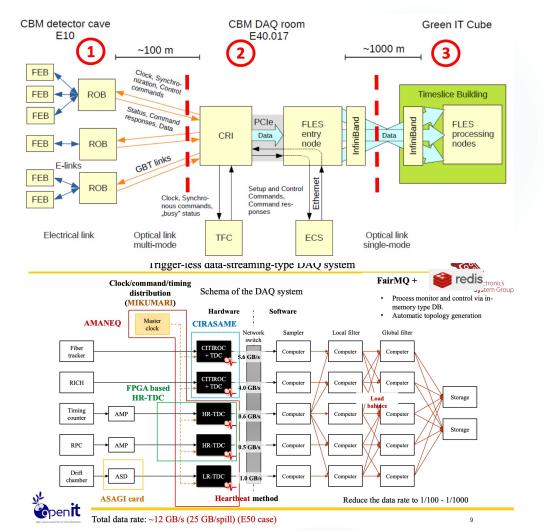


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Will be used in many experiments

ALICE, LHCb, sPHENIX, CBM, J-lab, J-PARC, etc...





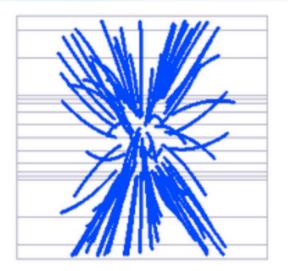
My activities: GPU processing

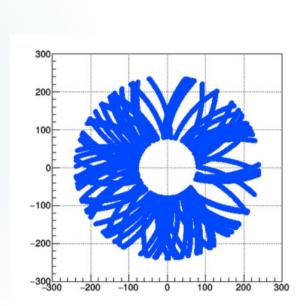
Kalman filter tracking

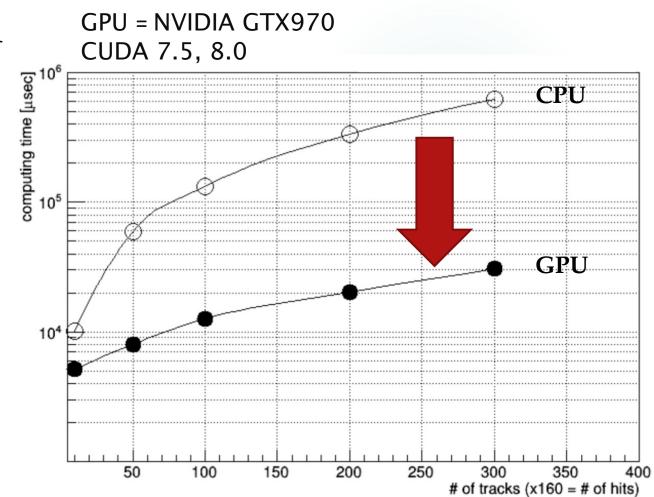
ML for pile-up identification

These works were done in 2016.

TPC Tracking





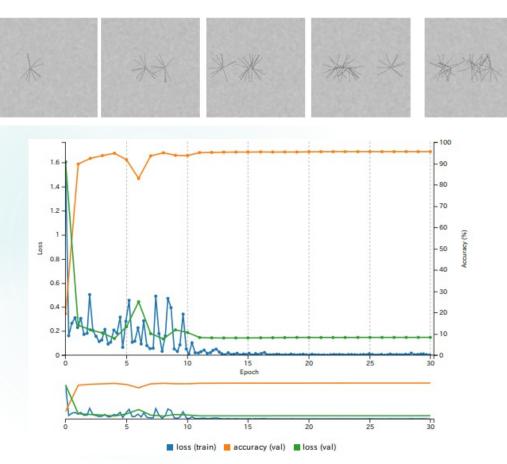


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My activities: GPU processing

Kalman filter in GPU

ML for pile-up identification



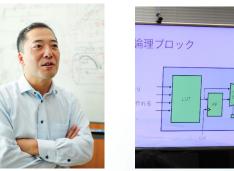
LeNET(vtxtmp) Image Classification Model

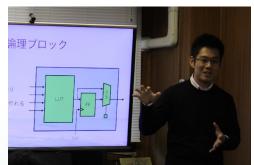
Path	Top pre	dictions	5							
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/1/0000.png	1 (100	0.0%	2	0.0%	3	0.0%	4	0.0%	5	0.0%
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/2/0000.png	2 100	0.0%	1	0.0%	3	0.0%	4	0.0%	5	0.0%
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/3/0000.png	3 99.	.99%	4	0.01%	2	0.0%	1	0.0%	5	0.0%
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/4/0000.png	4 99.	.99%	5	0.01%	3	0.0%	2	0.0%	1	0.0%
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/5/0000.png	5 100	0.0%	4	0.0%	1	0.0%	3	0.0%	2	0.0%
/home/gunjl/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/1/0001.png	1 100	0.0%	2	0.0%	3	0.0%	4	0.0%	5	0.0%
home/gunjliNVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/2/0001.png	2 100	0.0%	1	0.0%	3	0.0%	4	0.0%	5	0.0%
/home/gunjl/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/3/0001.png	3 00.	.92%	2	0.08%	4	0.0%	1	0.0%	5	0.0%
home/gunjl/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/4/0001.png	4 00.	.8%	5	0.2%	3	0.0%	2	0.0%	1	0.0%
homeigunjiNVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/5/0001.png	5 100	0.0%	4	0.0%	3	0.0%	1	0.0%	2	0.09
home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/1/0002.png	1 100	0.0%	2	0.0%	3	0.0%	4	0.0%	5	0.09
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/2/0002.png	2 100	0.0%	3	0.0%	1	0.0%	4	0.0%	5	0.0%
/home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/3/0002.png	3 (99,	.92%	2	0.08%	4	0.0%	1	0.0%	5	0.09
home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/4/0002.png	4 54.	.47%	5	45.53%	3	0.0%	1	0.0%	2	0.0%
home/gunji/NVIDIA_CUDA-7.5_Samples/deep_learning/tmp/vtxtmp/test/5/0002.png	5 99.	42%	4	0.58%	3	0.0%	1	0.0%	2	0.0%

TPC cluster finder in FPGA

ALICE TPC clustering on FPGA

- ▶ Xilinx Alveo u55c, u280
- High Level Synthesis (C++)

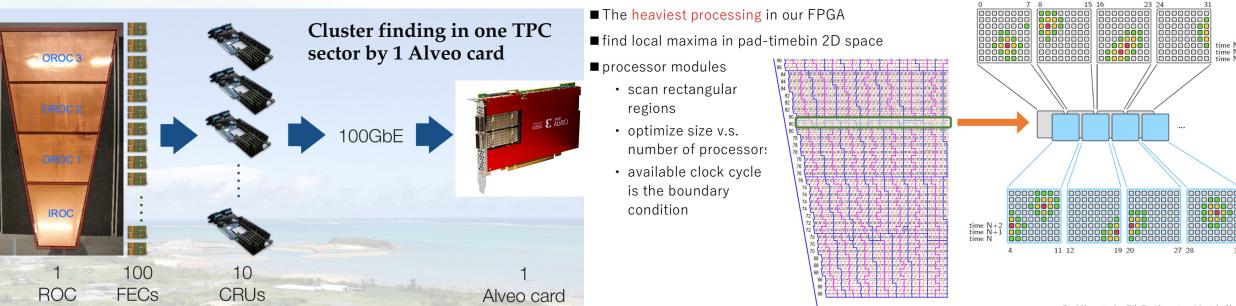




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Ken Oyama (NiAS)

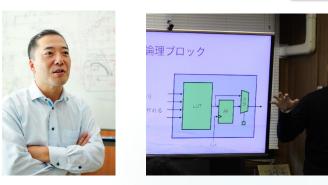
NiAS) Yasunori Osana (Rukyu -> Komamoto)



TPC cluster finder in FPGA

ALICE TPC clustering on FPGA

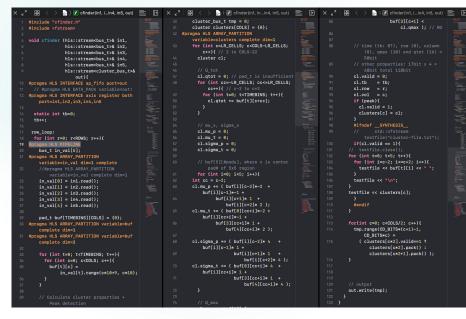
- ▶ Xilinx Alveo u55c, u280
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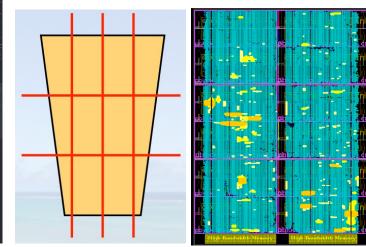


Ken Oyama (NiAS) Yasunori Osana (Rukyu -> Komamoto)

U55で10 CRU分が入った 290MHzで動作 → 5MSPSに対して58 clock

3	Used	Available	%
	714,307	1,303,680	54.8
FF	814,523	2,607,360	31.2
BRAM	846.5	2,016	42.0







Online tracking using ML



QF7 ---- QF5

- Ideal χ^2

BRAM

75 (2.8%)

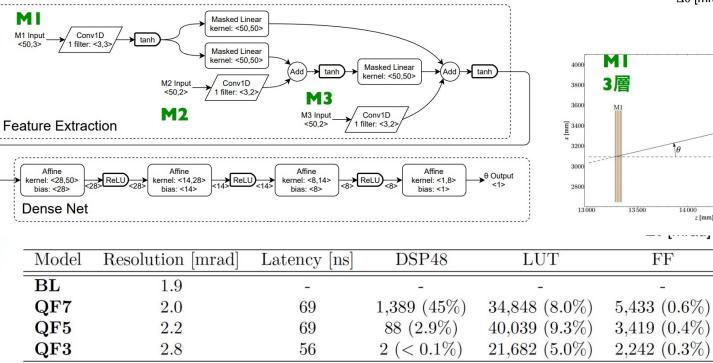
75 (2.8%)

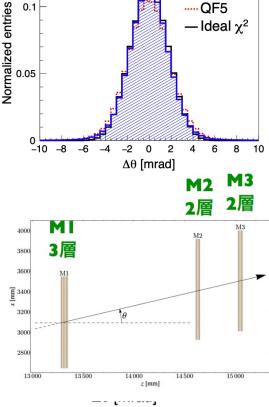
75(2.8%)

Prof. Horii (Nagoya Univ.) for ATLAS ATLAS muon tracking

- **Implementation of ML-based online tracking**
- Latency < 100ns







0.1

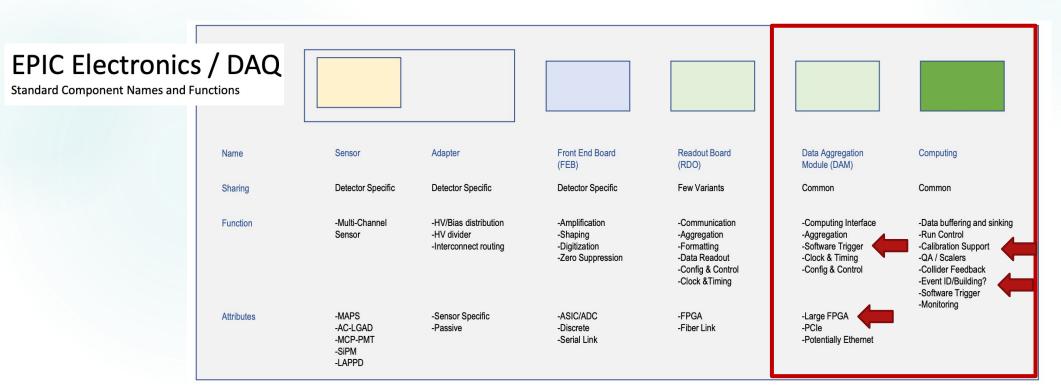
リソース使用率は、Super Logic Regionあたりの値

Interest in contributing ePIC-DAQ

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Plans for ePIC

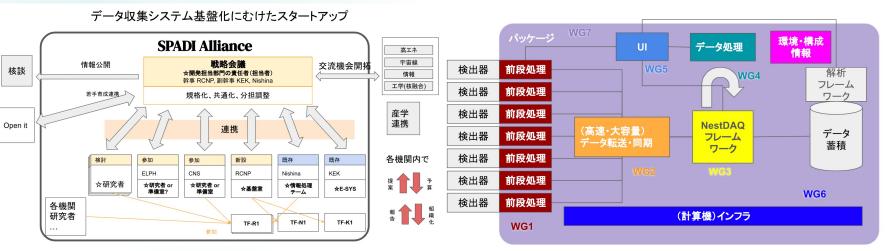
- Data processing and software trigger using hardware acceleration (FPGA, GPU, CPU)
- Vertexing and Tracking at central barrel, PID, jet reconstruction, etc...



SPADI-Alliance

"Signal Processing and Data Infrastructure – Alliance (SPADI-Alliance)" in Japan NP community to build and share common infrastructure for Streaming Readout at RIBF, J-PARC, etc...

- ASICs, FECs, streaming software, hardware acceleration
- https://www.rcnp.osaka-u.ac.jp/~spadi/
- ~80 members in total

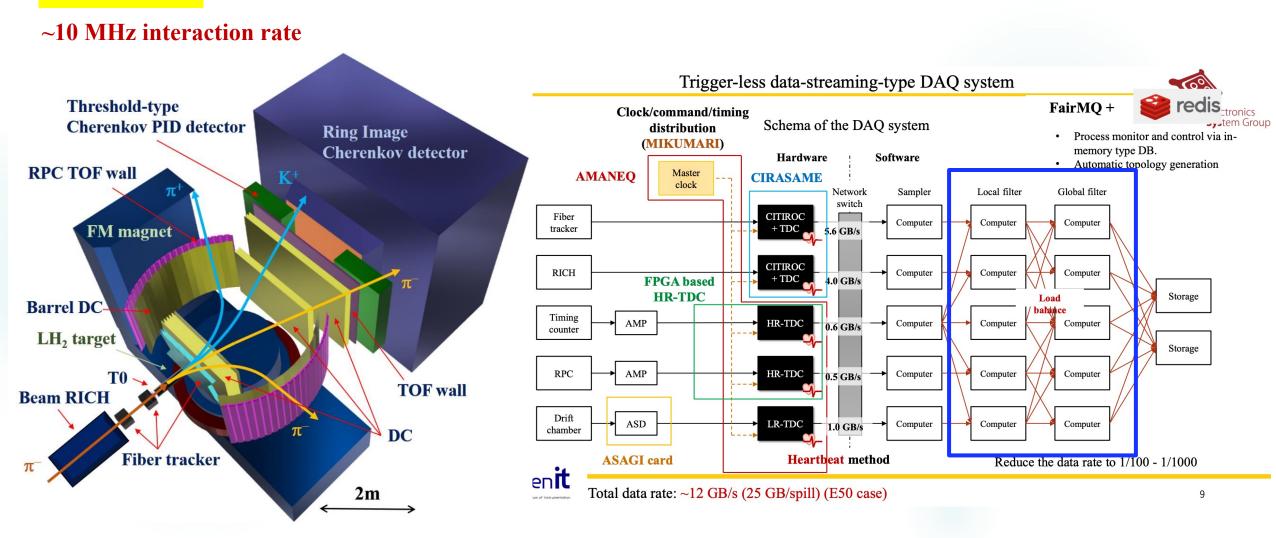


WG1. Analog (ASIC, Front-end) WG2. Data Transfer, Timing WG3. NestDAQ framework **WG4. Online processing (T. Gunji)** WG5. User Interface WG6. Infra-structure WG7. Packaging

Needs of online processing

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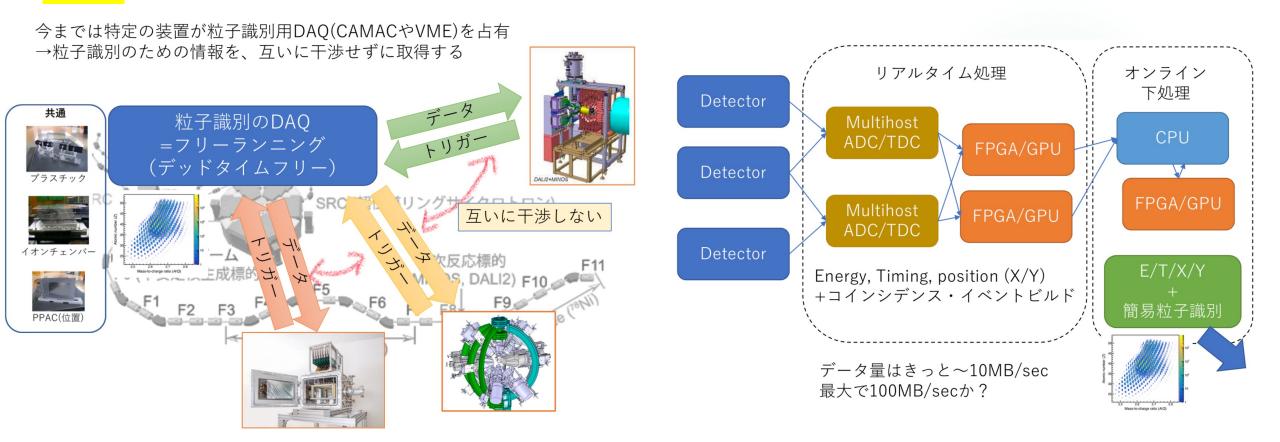
J-PARC E50



Needs of online processing

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RIBF



Interest in contributing DAQ

Plans for ePIC

Data processing and software trigger using hardware acceleration (FPGA, GPU, CPU)

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- Possible work items:
 - Run and get familiar with MC simulations
 - Benchmarking of:
 - online reconstruction (tracking, PID, etc)
 - event selections
 - usage of machine learning technique
 - Full system tests to evaluate computing models and estimate computing resources
- Close collaboration within SPADI-A.