

Real-time intelligent data processing for the next generation of particle imaging detectors



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Particle Imaging Detectors

Next Generation- Deep Underground Neutrino Experiment (DUNE)

Expected data rates: 5TB/s
Expected to be Operational by 2030

Current Generation- Short Baseline Near Detector (SBND)

Expected data rates: 45GB/s
Expected to be Operational by 2023

Past Generation- MicroBooster Neutrino Experiment (MicroBooNE)

Data rates: 33 GB/s
Data taking: 2015-2021

Liquid Argon Time Projection Chamber (LArTPC) detectors

Rich off-beam physics program

Excellent resolution

Rare physics processes: baryon number violation

Low energy blips

Data Processing Schemes

- ❖ For DUNE, processing > 5TB/s data would be computationally challenging
 - Requires intelligent data-processing and data-reduction schemes.

TPC-based trigger - Currently being demonstrated with MicroBooNE, soon with SBND

ADC count vs Time tick plot showing triggered and continuous streams with a channel dependent threshold.

Summary of ionisations per wire

Zero-suppressed waveform [2]

- ❖ The TPC-based trigger can be used to trigger on any off-beam activity including rare signals.
 - Future possibility includes triggering on some of the on-beam activities.

On-going Development and Demonstration

- ❖ Offline development and validation using MicroBooNE data.
- ❖ Developed trigger to tag Michel electron topology.
- ❖ Used **topological** (change in directionality) and **calorimetric** (energy in Bragg peak) information.
- ❖ 2% efficiency in selecting such events

Online Processing begins with generating drift regions (2.3ms Event) from frames (1.6 ms)

Following DUNE strategy [1]

Conclusions

- ❖ Online data selection is currently being demonstrated with MicroBooNE.
- ❖ Important proof-of-principle for upcoming SBND and future DUNE experiments

Future Possibilities

- ❖ Implement TPC-based trigger in SBND with a future goal of deploying in DUNE.
 - Utilize machine learning (ML) tools to target low and high energy activities with higher efficiency → demonstrated with DUNE simulated images of interactions

Uses CNN with VGG16 network architecture for image classification

Sample	Train Size	Test Size	Accuracy (%)	Inference Time (ms)
NB	12,023	4,027	99.53	0.47
LE	12,050	3,970	4.01	94.48
HE	10,137	3,417	3.63	6.15

~90% decision accuracy but high latency ~5 ms to run over 2.25 ms image.

Online inference in GPU using smaller than VGG16 network architecture → >3x improvement in inference time.

Fixed-point (QAT)	NB	LE	HE
true_NB	99.68%	0.32%	0%
true_LE	3.90%	94.69%	1.41%
true_HE	3.25%	6.35%	90.40%
total accuracy	95.16%		

[Y. Jwa, et al. arXiv:2201.05638]

- ❖ Possibility to use ML tools on specialized hardware such as FPGA (power efficient)

References

- ❖ [1] B. Abi et al., *JINST* **15** T08010 (2020).
- ❖ [2] P. Abratenko et al., *JINST* **16** P02008 (2021).
- ❖ [3] Y.J. Jwa, G. Guglielmo, L. Arnold, L. Carloni, G. Karagiorgi, *Front.Artif.Intell.* **5** (2022) 855184.