









Secretaría Nacional de Ciencia y Tecnología

High Event-Rate Online Discrimination on Mixed Radiation Fields with ML

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Summary

- Mixed radiation fields
- Event detection and discrimination
- High event-rate challenges
- Case study gamma/neutron
- Hardware setup
- Data collection and analysis
- Real-time event processing with ML
- Results and conclusions

Mixed radiation fields



Distribution of the energy deposited in the active volume of the Timepix3 Radiation Monitor by the particles composing a mixed radiation field at an LHC experiment [1].

Mixed radiation fields

- Ionizing radiation
- Chargeless and charged particles



Per-pixel energy deposited by various particles in a mixed radiation field measured using a Minipix Timepix detector with Silicone sensor [2].

Event detection



Captured trace from NaI(TI) detector placed nearby a gamma source [3].

Detector technologies

- Indirect charge collection
- Light detectors (with scintillators)
- Solid-state detectors
- Gas detectors









* digital domain starts at this stage in modern digital pulse processing (DPP) systems

Detection mechanisms

- Cross-level trigger (CLT)
- Constant fraction discriminator (CFD)
- Other advanced methods





- Different particle interactions -> different pulse shapes
- Real-time hardware deployment



Pulse shape discrimination (PSD) and frequency-based discrimination (FCI)



Pulse shape discrimination (PSD) and frequency-based discrimination (FCI)



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Feature extraction + machine learning



Feature extraction [6]

High-event rate challenges



Simulated pile-up events at high SNR

High-event rate challenges

- Pile-up distortion
- Baseline shifting
- Pulse shape for PSD or FCI distorted



Severe pile-up distortion and shifted baseline [5]

Pile-up rejection

- Severe pile-up:
 - Events discarded
 - Live-time reduced

- Detectors on current mode:
 - No discrimination



High-event rate + event discrimination

- Pulse shape analysis
- Feature extraction
- Machine learning model
- ML model compression
- Real-time deployment in hardware



Eight piled-up events from gamma/neutron detector [5]

Case study: γ / n discrimination



Gamma/neutron discrimination featuring novel frequency-based approach [8].

Why γ /n discrimination?

- Gamma radiation associated to neutron presence
- Several neutron applications in industry, medicine, energy, security, etc.







Hardware setup



Render of custom DAQ board for low-SWaP radiation monitoring system

Hardware setup

- IAEA/NSIL low-SWaP DAQ board
- Off-the-shelf CLYC detector







CLYC scintillator

- Triple-mode scintillator
 - Gamma spectroscopy
 - Better resolution than Nal(TI)
 - Similar sensitivity to Nal(Tl)
 - Thermal neutron detection
 - Enriched with Li-6 (neutron capture)
 - Fast neutron spectroscopy
 - Cl-35 neutron scattering
 - Optimized for gamma/neutron discrimination





Mixed radiation detector based on CLYC

- Integrated detector: compensated bias, output preamplifier
 - CLYC crystal: gamma rays, thermal neutrons, fast neutrons
 - SiPM sensor array: low SWaP (size, weight, power), magnetic field tolerance, robustness
- Higher output capacitance: challenging signal processing
- Reduced SNR and higher dead-time
 - Pulse length/duration: ~ 30 μs



Custom DAQ board



Custom DAQ board

- Low SWaP for portable applications
- FPGA for real-time signal processing
- Microcontroller (MCU) for peripherals
- 14-bit ADC @ 250 Msps
- Analog front-end (AFE)
- Programmable bias supply for SiPM
- Non-volatile flash: FPGA bitstream + detector data
- GNSS/GPS, RF interface, SiPM temp.



Data collection and analysis



Simulation of gamma/neutron discrimination using FFT with plastic scintillator [5].

Data collection

- Aiming to train supervised ML model
- Data recorded at NSF, IAEA with CLYC detector.
 - \circ AmBe source
 - Deuterium-deuterium gen.
- ~10^6 individual events recorded



Data wrangling

- Data curation with simplified correlation [8].
- Removed piled-up pulses
- Identified low-energy events



Data features after wrangling

- Slight pulse shape differences between gamma/neutron
- Diverse baseline shifts
- Sampled at 4 Gsps
 - Further downsampled for real-time processing
- No pile-up distortion



Data tagging

- Frequency-based event discrimination [9].
- Two labels: gamma/neutron





Pile-up synthesis for ML training

- Exponential distribution of "clean" piled-up pulses
- Event rate at 200 kHz (max 400 kHz)
- Validated event time distance with R² ~ 1.00



Real-time event processing with ML



Simulation of gamma/neutron discrimination using FFT with plastic scintillator [5].

FPGA system architecture

- Low/high-power domains: 100/200 MHz
- Real-time processing pipeline: II = 995 ns, latency < 2.5 µs



Real-time feature extraction

- Pulse leading edge
- First 350 ns (pulse 30 µs)
- No time alignment required
- 2nd derivative trigger (SSD)
 - IIR bandpass differentiator
- Python model
- VHDL real-time module



Real-time feature extraction



Machine learning workflow



Online machine learning classification

- Multilayer perceptron (MLP): binary classifier.
- Compression workflow from [9].
- Distillation + quantization-aware pruning
- 8-bit FP quantization with 30% sparsity.
- 217 parameters in 6 hidden layers.
- Overall accuracy 98.2%.



Results

Results: performance

- Count-rate (CR)
- Inverse of dead-time (1/DT)
- Pile-up recovery/rejection (PuP)
- Accuracy (Acc)

Work	Overall perf.		
This work	1.0		
Michels et al.	0.89		
Wen et al.	0.12		
Cruz et al.	0.01		
Astrain et al.	0.01		



Highest values are the best

Results: low SWaP

SWaP comparison Thermo Fisher Scientific https://www.thermofisher.com/order/catalo g/product/4250631 P (W) Hardgrove et al. doi:10.1109/MAES.2019.2950747 Mesick et al. W (kg) https://digitalcommons.usu.edu/smallsat/20 20/all2020/118/ Thermofisher Zhao et al. Hardgrove doi:10.1088/1748-0221/18/09/P09043 Mesick S (L) Zhao Huang Huang et al. This work doi:10.2139/ssrn.4717223 3 12 15 6 9 0

Lowest values are the best

Results: FPGA utilization

- Low-end Artix-7 35T FPGA
 - LUT: 30.4%
 - Registers: 17.4%
 - BRAMs: 24.0%
 - DSPs: 17.8%



Validation with γ +n events

- Recorded pulses from CLYC plugged into AFE
 - (a) Gamma events
 - (b) Neutron events
- Runs
 - (i) Gamma-only events
 - (ii) AmBe + Cs-137 sources
 - (iii) DD neutron gen. + Cs-137
- Neutron false alarms < 2.5%
- Accuracy: 98.2%



Validation with other detectors

- Nal(Tl) detector sensing γ events from Co-60 source
 - Accuracy 99.1%
- Synthetic γ events of fast plastic scintillator
 - Absolute max. countrate:
 ~1.01 Mcps
 - Lowest deadtime: 995 ns
 - 129 ppm (missing events)



Conclusions



Experimental gamma/neutron dataset after discrimination with frequency-based analysis for ML training

Conclusions and FW

- FPGA enables a flexible mixed-radiation detection and measurement platform
- Reliable event discrimination under pile-up distortion on mixed radiation fields has been achieved
- Low-SWaP instrument with benchtop performance enabled by ML
- Multi-detector systems might be deployed leveraging existing system architecture
- Targets include portable instruments in nuclear security, radiation monitoring, and HEP experiments.

References

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Thanks for your attention



MLab, ICTP (Italy)

ECFM, USAC (Guatemala)

Backup slides



SWaP comparison

Work	S (L)	W (kg)	P (W)	SaW	WaP
This work	0.56	0.41	1.5	0.13	1.00
Huang 153	4.08	4.9	8.8	0.00	0.01
Zhao [18]	3.34	-	3.5	-	-
¹ Mesick [221]	-	7.0	14	-	0.01
¹ Hardgrove [106]	-	3.4	9.6	-	0.02
Thermofischer [105]	0.18	0.16	-	1.00	-



Table 5.2: SWaP comparison of recent CLYC-based γ/n discrimination systems, including SaW and WaP scores

PSD vs FCI: gamma-only dataset



Dataset FOM for ML training



Related applications



Remote diagnostics

