Scintillating Bubble Chamber Silicon Photomultiplier Analysis

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Scintillating Bubble Chamber:

SBC-LAr10

- Quasi-background-free Bubble Chamber
- 10 kg of liquid argon
- Aims to hit lower thresholds (100 eV)



How does SBC work?

- Relies on electron recoils to lose energy through scintillation
- Bubbles nucleate after small energy depositions
- SiPMs detect scintillation light to veto high energy singlebubble events





Silicon Photomultipliers (SiPMs)

- Detect single photons near UV to near IR
- SBC SiPMs detect LAr and LXe UV light
- Consists of many single-photon avalanche photodiodes (microcells)

Understanding SiPMs

- A microcell drifts towards an avalanche region when a photon hits
- The microcell then activates a self-sustaining avalanche
- **Dark Event:** generation of pulse without light
 - Crosstalk
 - After Pulsing





SiPMs Background Crosstalk

- Occurs when primary avalanche in one microcell triggers another in an adjacent cell
- Occurs in on top or near initial pulse
- Causes increased initial pulse magnitude
- May have inflection point



SiPMs Background After-Pulsing

- There is a probability that an adjacent cell may release a delayed pulse
- Occurs in recover phase
- Lesser magnitude than original if near original

Motivation

To explore new methods of determining the dark noise rate and cross talk probability using machine learning methodologies and compare to algorithmic approaches.

Pulse Counting with Computational Algorithms



VUV-sensitive Silicon Photomultipliers for Xenon Scintillation Light Detection in nEXO

Jamil et al. 2019



Unsupervised Machine Learning

K-means Clustering

50 Clusters • 10 000 Initial Pulses





Cluster 12

After Pulsing 31 Pulses



Cluster 1 Delayed Crosstalk 160 Pulses

Resulting Algorithm





Supervised Machine Learning

~4000 hand-picked labels

Algorithm
Machine Learning



Current Results

Traditional Algorithms

- Accurately calculates t0
- Calculates charge
- Near 100% accuracy on simulated data

Machine Learning

- Real Dataset
 - Near-even distribution of labels
 - ~80% accuracy
- Trial run
 - 1000 unseen datapoints
 - Uneven distribution of labels
 - ~90% Accuracy

Conclusion

- Unsupervised ML effectively classifies background groups
- Supervised ML can be used for pulse analysis
- ML can effectively improve current algorithmic approaches



Thank you

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