CND Neutron Detection

Erin Seroka CLAS Collaboration Meeting March 23, 2023 Paris, France

Overview



- Neutron Detection efficiency
- Development of neutron veto algorithm
 - Early stages
 - Development of ML with
- ML neutron detection with RGM data

Detection Efficiency: early work



Detection Efficiency: Approach

- Channels used: $h(e, e'\pi^+n)$, $d(e, e'p_{CD}n)$, $d(e, e'p_{FD}n)$
- Background subtraction



Detection Efficiency: results



Neutron Veto

- CND neutron reconstruction
 - Clusters form neutral seed if unassociated with a CVT track
 - Neutrals considered to be only photons or neutrons
 - Velocity cut: $\beta < 0.8$ for neutron, $\beta > 0.8$ for photon
- Big problem: imperfect CVT efficiency means proton contamination
- Background sources
 - Double hits
 - Neutron and proton reconstructed in same place (cut on θ_{np})
 - Random co-incidence (off-time, etc.)
 - Charged particles
- Past work
 - Andrew Denniston: preliminary CND veto work
 - Adam Hobart: Machine learning for DVCS

Background: double hits

- Charged particles may leave two hits with two sets of PMT signals
- Two PMT signals arrive at same time -> reconstructed near 40°
- Easily eliminated using z cut



Background: CND hit not associated w/ track

- CND hits not associated with CVT track
- Proton associated with one CND cluster but not another
- Can cut on angle between neutron and proton



Background: protons misidentified as neutrons

- Imperfect CVT tracking efficiency
- Some protons misreconstructed as neutrons
- This is the main background source we seek to eliminate with ML



Neutron Veto: Development in Simulation

- Approach using Machine Learning: Boosted Decision Trees
 - Identify features that are best at distinguishing between real neutrons and "fake" neutrons (non-neutrons with neutron PID)
- Goal for features
 - Local, detector-level information
 - Avoid kinematic variables

Neutron Veto: Development in Simulation

- Sample generation
 - Uniform *e'n* and *e'p* generators with nucleon momentum up to 1 GeV/c
 - Run through GEMC, generated momentum "truth" preserved
 - Added CLAS12 background from RG-A
- Good neutron sample (signal): agreement with generated momentum in e'n+bknd
- Fake neutron sample (background): all neutron PID in e'p+bknd

Neutron Veto: simulation feature list

- Number of hits in 5 CND sectors closest to neutron
- Energy deposition in 5 CND sectors closest to neutron
- Number of hits in CND cluster
- Neutron energy
- CND layer multiplicity
- Number of hits in 6 CTOF components closest to neutron
- Energy deposition in 6 CTOF components closest to neutron



Neutron Veto: sim feature list









Neutron veto: simulation results



	feature	importance
3	hits_nearby7	0.400353
5	CTOF energy nearby	0.398342
4	avg energy per hit7	0.084178
1	layermult	0.055034
0	n energy	0.050658
6	CTOF hits nearby	0.006834
2	size	0.004600

Neutron Veto: data approach

- Available exclusive channels
 - $h(e, e'\pi^+n)$
 - $d(e, e'p_{CD}n)$ (2 GeV, 6 GeV)
 - $d(e, e'p_{FD}n)$ (2 GeV, 6 GeV)
 - $d(e, e'pp\pi^-)$
- Select good and false neutrons
 - Start with $d(e, e'p_{CD}n)$ QE channel (higher stats)
 - Calculate expected neutron momentum with momentum conservation
 - Train ML to separate neutrons vs bad neutrons

Neutron Veto: good neutrons (signal)

- $\cos \theta_{n,pred} < 0.8$
- | p_{miss} p_n | > 0.2 GeV/c
- M_{miss} > 1.15 GeV/c



Neutron





Neutron Veto: non-neutrons (background)

- $\cos \theta_{n,pred} > 0.9$
- $| p_{miss} p_n | < 0.1 \text{ GeV/c}$
- M_{miss} < 1.05 GeV/c











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Neutron Veto: Results



	feature	importance
6	CTOF nearby hits	0.558750
2	size	0.157170
3	CND nearby hits	0.100631
5	CTOF nearby energy	0.056925
0	neutron energy	0.046822
1	CND layer multiplicity	0.043617
4	CND nearby energy	0.036085

Next Steps

- Detection Efficiency
 - Re-run with pass2 reconstruction
 - CLAS analysis note
- Neutron Veto ML Algorithm
 - Channel to focus on protons mis-reconstructed as neutrons (e.g. $d(e, e'pp\pi^{-})$)
 - Continue search for good features (e.g. number of nearby hits in CVT)
 - Cross-tests on different data sets