

Neural Network for Neck Discrimination

Light Pattern Filtered by Time Pattern

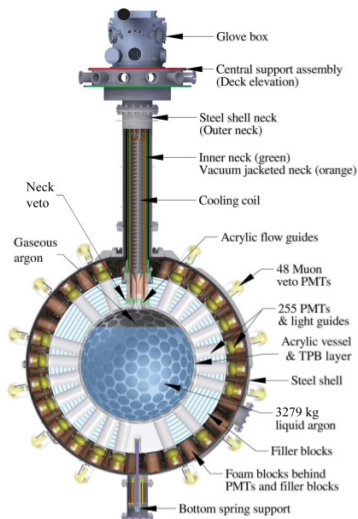
CIEMAT Team

CIEMAT

Madrid, 20 octubre 2022

DEAP-3600 detector

- Placed in SNOLAB under 2 km of rock (≈ 6000 mwe of coverage).
- Spherical acrylic vessel with diameter of 170 cm. It is filled with 3279 kg Liquid Argon.
- Single phase detector with 255 PMTs covering the 75% of the sphere. Grouped in 35 rings from 5 to 10 PMTs.



DEAP-3600 analysis

- WIMP dark matter search with 2016-2020 dataset
 - Profile likelihood ratio analysis (open data strategy)
 - Blind analysis on full dataset
- Other searches
 - 5.5 MeV solar axions
 - ${}^8\text{B}$ neutrino absorption signal (inverse beta decay)
- ${}^{39}\text{Ar}$ Measurements
 - ${}^{39}\text{Ar}$ specific activity
 - ${}^{39}\text{Ar}$ half-life
 - ${}^{39}\text{Ar}$ decay spectrum and nuclear parameters
- Muon veto instrumentation and muon flux measurement
 - Muon flux annual modulation (or absence thereof)

The Problem

- Neck events discrimination (noise) vs. nuclear recoils (signal).

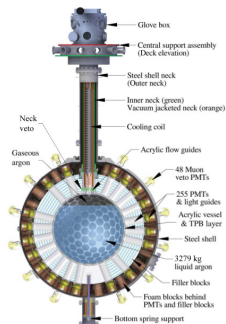
Only light pattern. Rejection 99.9%

pre-AI Rejection 99.0%, $A \approx 18$

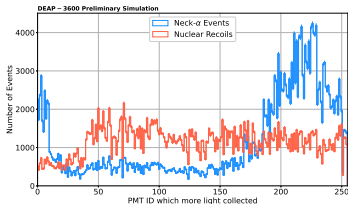
2019 MLP: D256D32 (100k events per label) $A = 41.5$,

2020 MLP: (300k events per label) $A = 44.7$,

2020 Inception (300k) $A = 57.9$,
Inception = CNN.

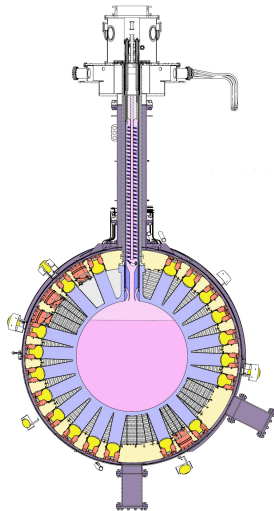


Patterns

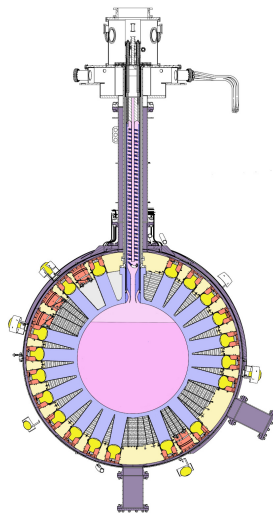
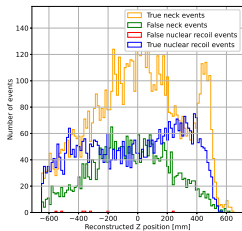
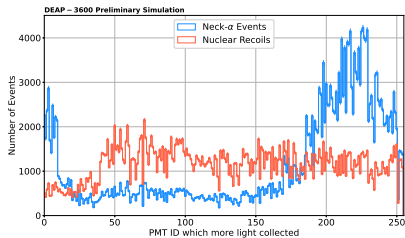


Incorporate time pattern

To optimize the preprocessing of data (dividing the light pattern in windows of time) based on parameters (-Acceptance) obtained from other process (NN discriminating neck events).

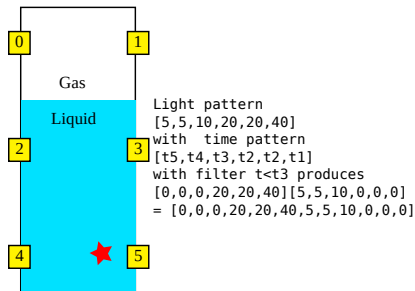


Patterns



Idea

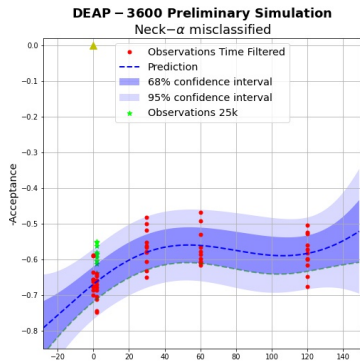
- To use the time pattern with time windows for filtering the light pattern.
- Low and high end times at -28 and 1000 ns. Optimization of the time divisor needed: $(-28, x)(x, 1000)$. Since 1 time divisor is used, the light pattern filtered has the double of the length of the original one (255X2).



- 1D input. Lost information in Z coordinate.

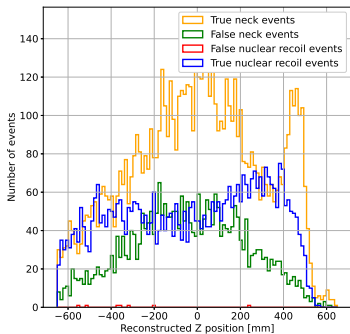
Bayesian Optimization of Time Divisor

- Bayesian optimization of time divisor. 2 time windows created: $(-28, \text{time-divisor})$ and $(\text{time-divisor}, 1000)$ ns. 11 runs.
- Minimization of $-Acceptance$ in the discriminator of neck events. The lower value, the better acceptance.
- 5 time divisor tested: 0, **2** (best), 30, 60 and 120 ns.

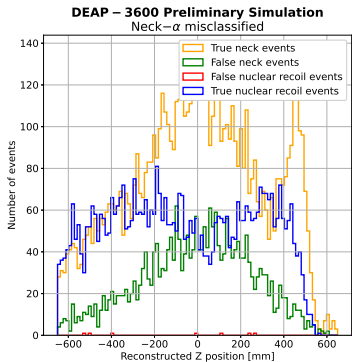


With the whole dataset (imbalanced) the mean Acceptance is (25k per label) **0.686** and **0.587** without filtering.

Time divisor off-on Z coordinate. 25k events/label



(a) off, $A=0.620$



(b) on, $A=0.768$

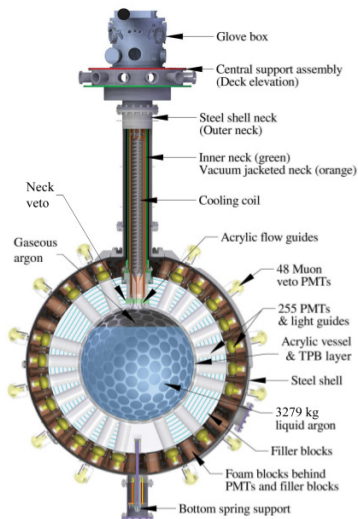
Why?

Toy model/explanation

- $\frac{1}{v_g} = 7.5 \pm 0.007 \pm 0.05(\text{sys})$
ns/m, $\lambda = 128$ nm.

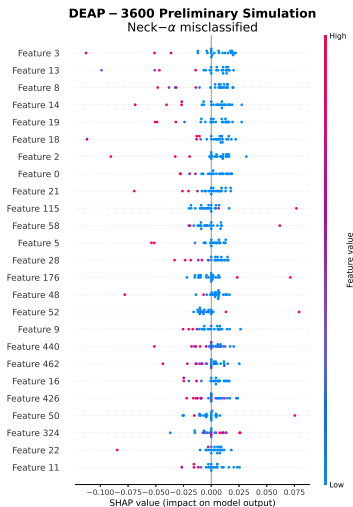
Propagation of scintillation light
in Liquid Argon: M. Babicz et al
2020 JINST 15 C03035

- $v_g \times 2\text{ns} = 0.266$ m,
 $v_g \times 3\text{ns} = 0.400$ m from -1 ns
to 2 ns ?? related to acrylic
vessel of radius 0.850 m.



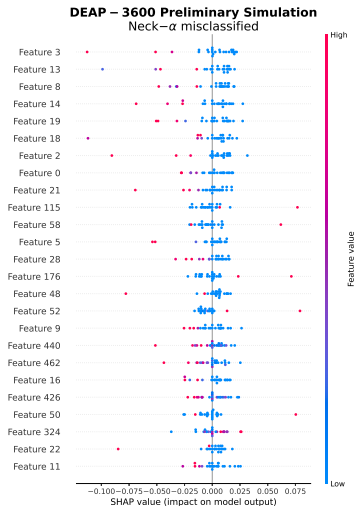
XAI: SHAP

- Analysis for 25 events (memory exhausted by Shap).
- The first window (-28,2) ns of the top PMT are critical for the classification: 3, 13, 8, 14...
- Exception in features 440 (second time-window of PMT 185), 462 (207), 426 (171), 324 (69), all in liquid.
- Larger number of blue dots (no/low light) than red ones because the sparse light pattern.



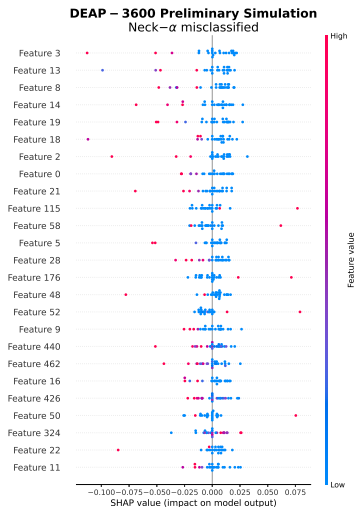
XAI: SHAP

- Not red points in both sides of shap values a single feature. Strong relevance for the classification.
- Strongly negatively correlated: low values of variables (blue dots) (low or not light in PMTs) slightly increase the shap value (moves the classification toward the class 1).
- Exception in few features (52, equatorial PMT) with positive correlation.

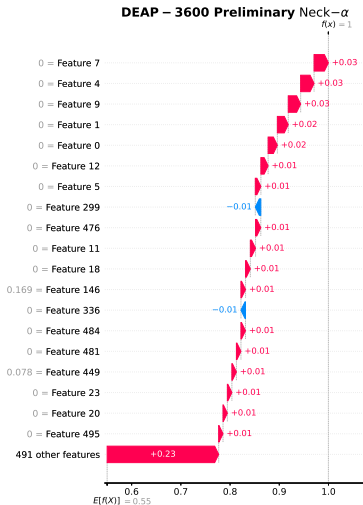
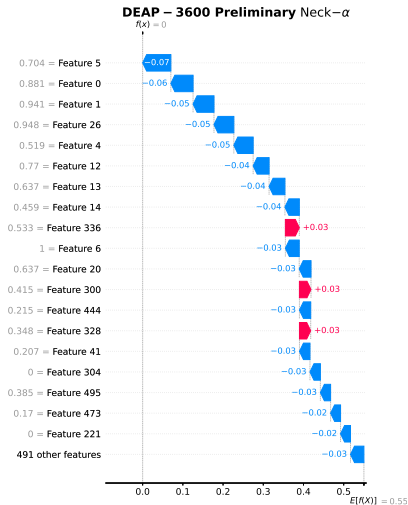


XAI: SHAP

- Few cases with high values (red dots) (a lot of light) in PMTs of low index drive to reduce the shap value (class 0).
- Oppositely in equatorial PMTs (115, 176) increase the shap value (move them to class 1).



XAI: SHAP



Skills

- Deep Learning under Particle Physics vision, where Explainability plays a key role.
- Adaptability to requirements: format input, most suitable architecture, custom loss, problem category, etc.