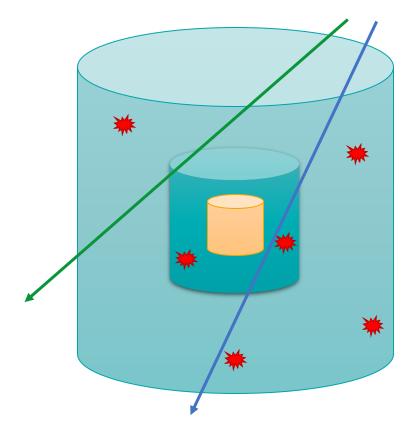
COSMOGENIC REJECTION STUDIES

(meeting 30.07.2020)

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Brief recap

- Goal: "smart" cosmogenic rejection from LAr veto acquisitions (no active water muon veto, no Ge coincidence):
 - <u>Target</u>: muon-induced processes (*Ionization*, *Compton*, *Ar41 de-excitation*)
 - Other <u>Background</u>: Beta-decays of Ar39
- Muon Events:
 - High-Detection Muons: "easy" task => Cut on Number of PE Detected
 - Low-Detection Muons (or Low-Energy Muons):
 - Muons that marginally cross the sensitive region
 - Resulting in a small amount of energy far from the fiber shrouds
 - Are they dangerous as well?
- Why Ar39 is a problem?
 - <u>High-activity</u> (1.41Bq/l) and integration over time (10us)
 - would result in <u>Pileup</u> of Ar39 decays
- Classification only relying on LAr:
 - **Descriptors**: amount of **energy** detected and **spatial distribution** of detections.
 - **Prompt Classification**: detections within O(10us) from trigger
 - **Delayed Classification**: correlate later detections with delayed processes in LAr (*Ar41 de-excitation*)
- Previous presentation: https://indico.legend-exp.org/event/341/contributions/1800/attachments/1106/1605/ML4NP_SimGroup_20200611.pdf

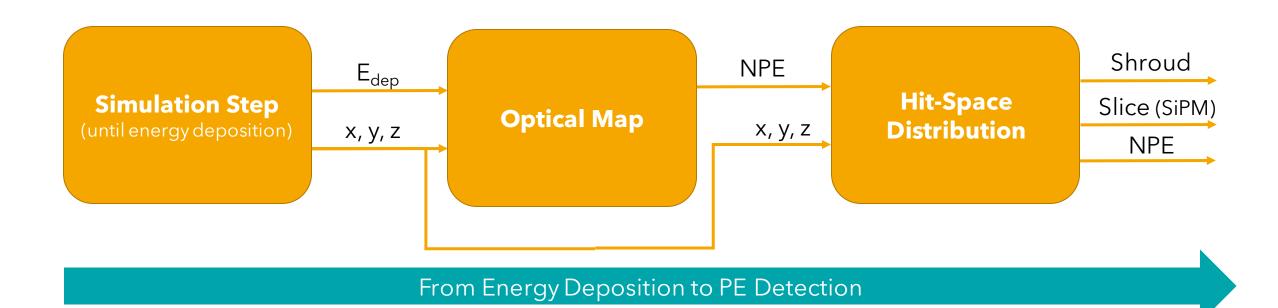


Updates since last talk (11.06.2020)

- Emulation of SiPM readouts:
 - Hit Space Distribution: Toy Spatial Map
- Extended Simulation in LGND200:
 - **3M Muons** (CJ):
 - LNGS conditions: input from MUSUN + 3m rock surrounding the volume (showering)
 - 10M Ar39 (CJ):
 - Primary particles are generated randomly in LAr volume
 - 2M Neutrons (Danila):
 - Starting Energy from Muons' Simulation
 - Position of primaries sampled in a cylinder wt H=4m, R=2m (LAr volume)
 - Theta in [0, 180), Phi in [0, 360)
- Prompt Classification with Machine Learning

Simulation Workflow

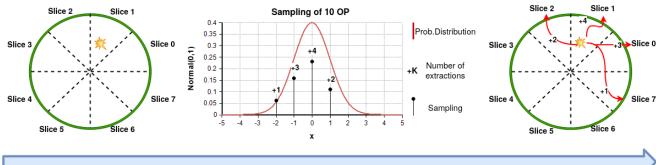
• To simulate detections in LAr veto, without optical simulation.



Hit Space Distribution

• Last meeting: we use a constant Gaussian distribution to sample the spread of

detection over the shroud.



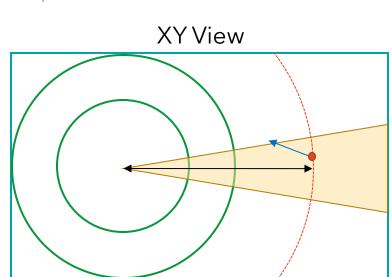
From Energy Deposition to OP Detection

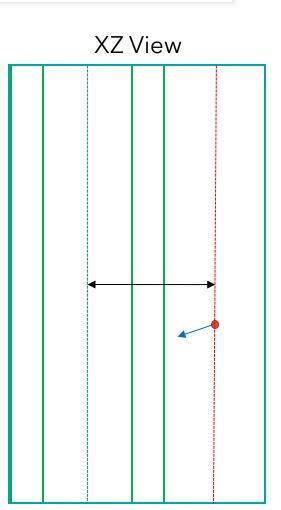
Problems:

- The **radius** of photon production point strongly affects the detection spread.
- LGND200 has 2 shrouds with 12 and 20 SiPM modules.
- Idea: use geometry!
 - Photon trajectory = straight line, Shrouds = circles
 - From the intersection of a line with circle, we can derive the hit-angle distribution
- No optical simulation (no WLS, Radon Shroud reflections), just a Toy Spatial Map

Toy Spatial Map (1)

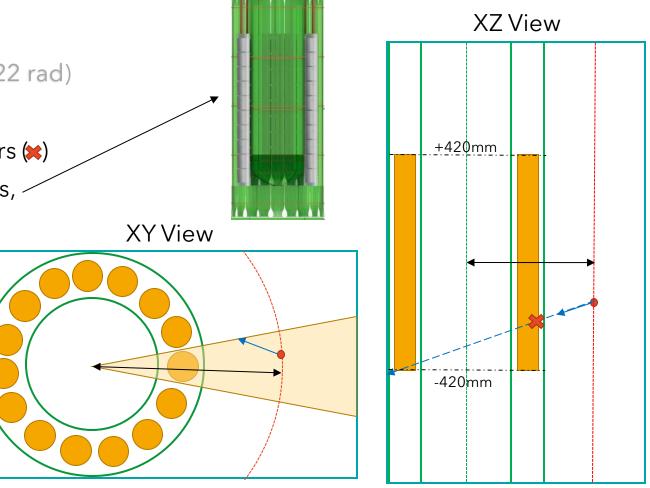
- For each **r** in [0, 700]:
 - Sample x, y, z s.t. radius is r in XY plane (+- 0.22 rad)
 - Sample OP trajectory
 - See where OP trajectory intersect Ge cylinder
 - If intersection in +-425: enable 14 Ge Crystals, otherwise no Ge
 - Compute all possible intersections with inner and outer shrouds
 - Sample intersection according to Pr
 - Compute angle of hit





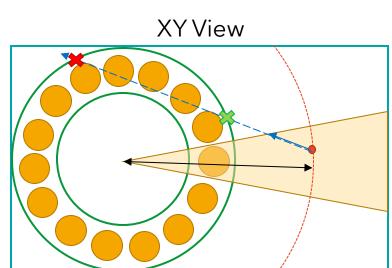
Toy Spatial Map (2)

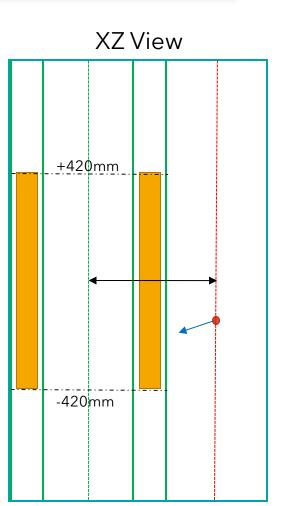
- For each **r** in [0, 700]:
 - Sample x, y, z s.t. radius is r in XY plane (+- 0.22 rad)
 - Sample OP trajectory
 - See where OP trajectory intersect Ge cylinders (*)
 - If intersection in +-420: enable 14 Ge Crystals, otherwise no Ge
 - Compute all possible intersections with inner and outer shrouds
 - Sample intersection according to Pr
 - Compute angle of hit



Toy Spatial Map (3)

- For each **r** in [0, 700]:
 - Sample x, y, z s.t. radius is r in XY plane (+- 0.22 rad)
 - Sample OP trajectory
 - See where OP trajectory intersect Ge cylinders (*)
 - If intersection in +-420: enable 14 Ge Crystals, otherwise no Ge
 - Compute all possible intersections with inner and outer shrouds
 - Sample intersection according to Pr
 - Compute angle of hit



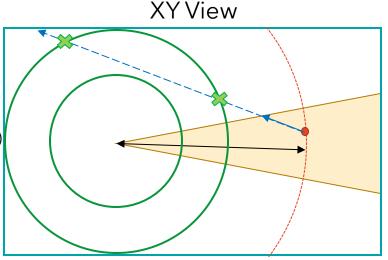


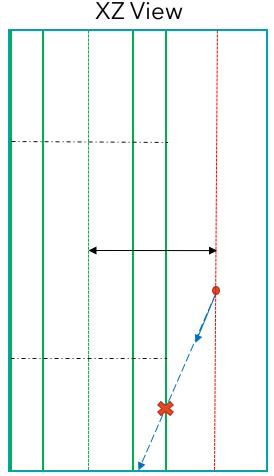
Toy Spatial Map (4)

P(geom.cov.) := shroud spatial coverage ~ 0.54

- For each **r** in [0, 700]:
 - Sample x, y, z s.t. radius is r in XY plane (+- 0.22 rad)
 - Sample OP trajectory
 - See where OP trajectory intersect Ge cylinders (x)
 - If intersection in +-420: enable 14 Ge Crystals, otherwise no Ge
 - Compute <u>all possible intersections</u> with inner and outer shrouds
 - Sample intersection according to Pr
 - First hit: P(geom. cov.) * $P(\lambda_{att})$
 - Second hit: (1-P(geom.cov)) * P(geom.cov) * P(λ_{att})
 - ...
 - Compute angle of hit

Another example (Ge Disabled)

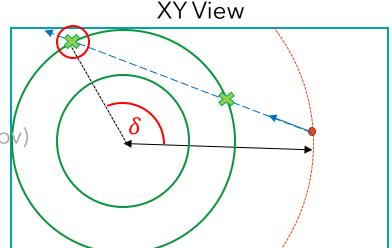


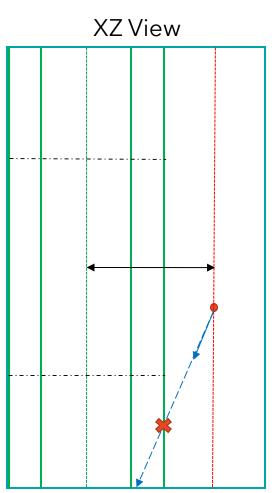


Toy Spatial Map (5)

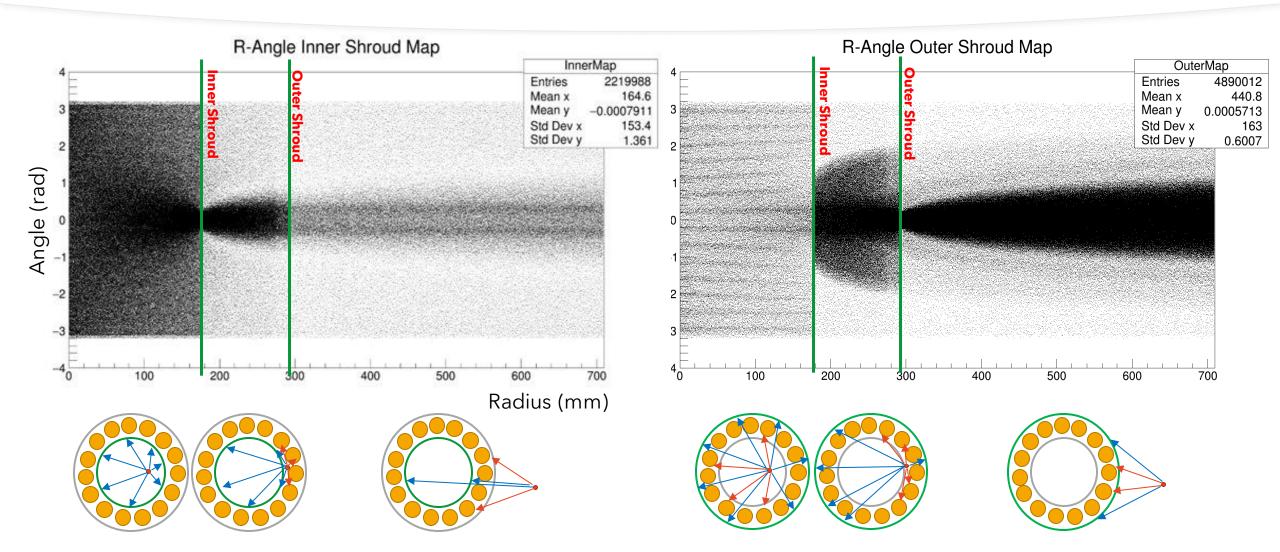
- For each **r** in [0, 700]:
 - Sample x, y, z s.t. radius is r in XY plane (+- 0.22 rad)
 - Sample OP trajectory
 - See where OP trajectory intersect Ge cylinders (**)
 - If intersection in +-420: enable 14 Ge Crystals, otherwise no Ge
 - Compute <u>all possible intersections</u> with inner and outer shrouds
 - Sample intersection according to Pr
 - First hit: P(geom. cov.)
 - Second hit: (1-P(geom.cov)) * P(geom.cov)
 - •
 - Compute angle of hit

Another example (Ge Disabled)





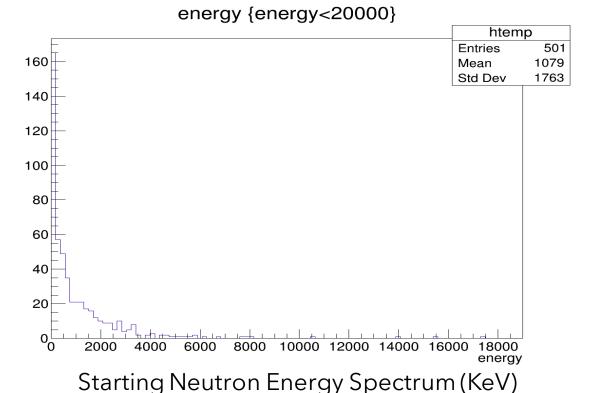
Toy Spatial Map - Result



Neutron Simulation

We extract the initial energies of neutrons from Muons Simulation to create an **Energy Spectrum**

We have ~20K events with Ar41 de-excitations, over 2M neutron simulations (1%)

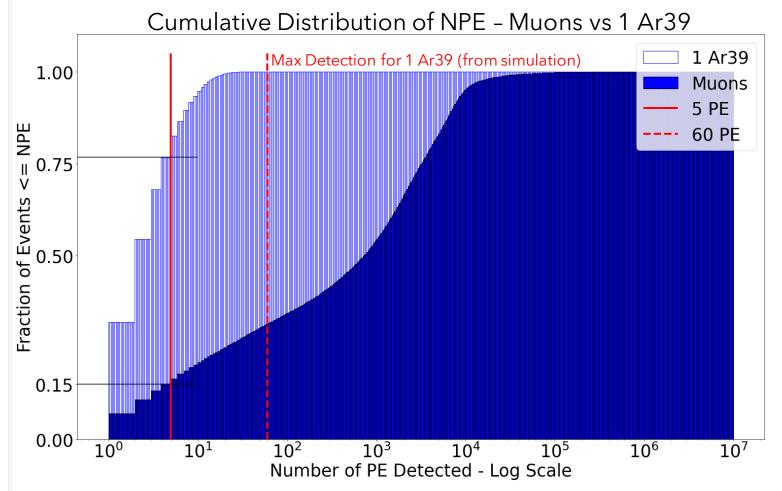


gamma energy h1 22690 Entries Mean 5.248 3500 2.147 Std Dev 3000 2500 2000 1500 1000 500 As expected, high-energy gammas production by Ar41 de-excitation

Post-processing is still on going...

Machine Learning - Prompt Classification (1)

- Simulation Parameters: Light Yield=40 OP/KeV, Q.E.=0.40
- First Problem: Single Ar39 decays
 - E[Ar39 Rate|1.41 Bq/l, V=2.6m3] = **3666 Hz**
 - From Simulation: ~65% Ar39 events wt NPE>=1
 - E[Ar39 Det. Rate] = 3666 * .65 = **2353 Hz**
 - Pr[>=1 Ar39 decays] = 1-Pois(k=0, rate) = .023255
 - Pr[1 Ar39] = Pois(k=1, rate) = **.022983**
 - Single Ar39 decays is dominant => 98.83 %
- Preliminary Cut: NPE>=5
 - From Simulation: ~75% of 1 Ar39 has NPE<5
 - From Simulation: ~15% of Muons has NPE<5

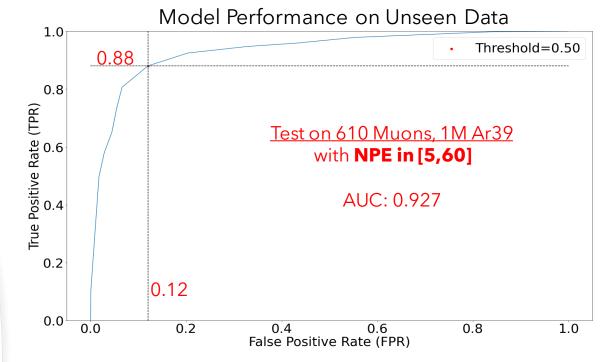


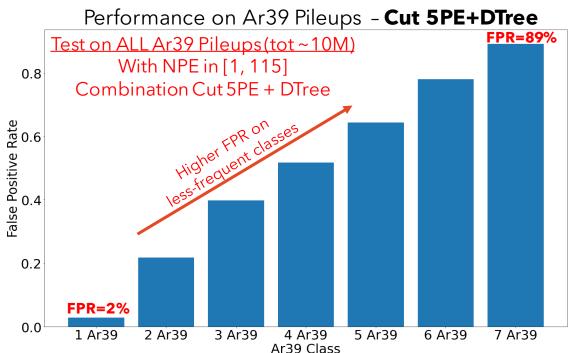
Including events NPE<5 PE, tough classification => Learning get worse Since Ar39 has a dominant rate compared to Muons => Classify NPE<5 as Ar39

Prompt Classification (DTree)

- First Problem: Single Ar39 decays
- From Simulation: Ar39 have NPE between 0 to 60 PE
- Preliminary Cut: NPE>=5
- Specialized Model "Mu vs 1Ar39":
 - Fast inference: Decision Tree
 - Simple Features for each shroud: NPE, Number of Active Slices
 - Explainable Logic
- Training Phase:
 - Events wt NPE in [5, 60]
 - Training Data: 4500 muons, 4500 Single Ar39

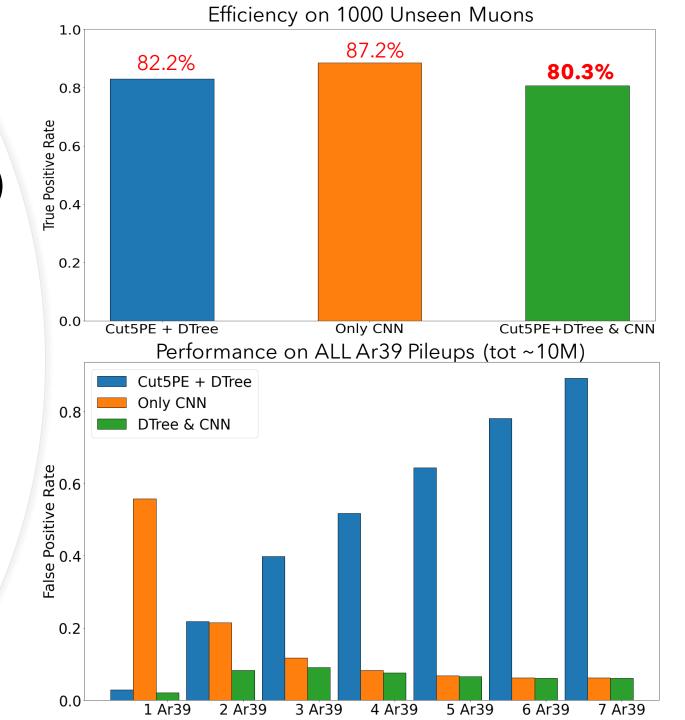
...how to deal with Ar39 pileups?





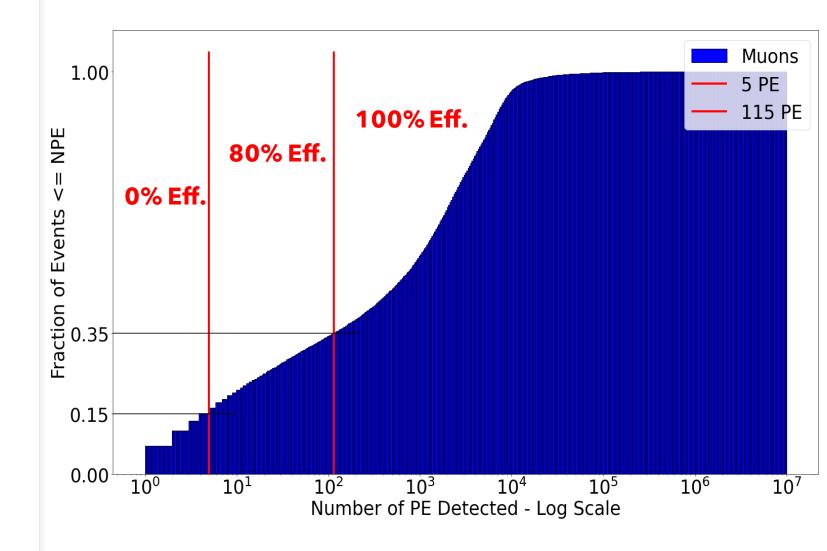
Prompt Classification (CNN)

- How to deal with Ar39 pileups? **2-Steps Approach**
- General "Mu vs Ar39 Pileups":
 - CNN Model
 - Train a model on events wt 5<=NPE<=115
- Combination of DTree and CNN:
 - Still matter of investigation...
 - Example: Dtree & CNN
 - Test on Muon Sample: only 2% Loss in Efficiency
 - Significant reduction of FPR



Prompt Classification - Recap

- In this simulated context...
 (OP Yield, Q.E., Optical Map, Toy Spatial Map)
- Classification Strategy:
 - Below 5 PE: Ar39
 - Between 5 and 115 PE: DTree & CNN
 - Above 115 PE: Muons
- Its efficiency:
 - Below 5 PE: 0%
 - Between 5 and 115 PE: ~80% on test set
 - Above 115 PE: 100%
- Overall expected efficiency:
 - $0.15*0 + 0.20*0.80 + 0.65*1 \sim 81\%$
- ...a great challenge is on purity!



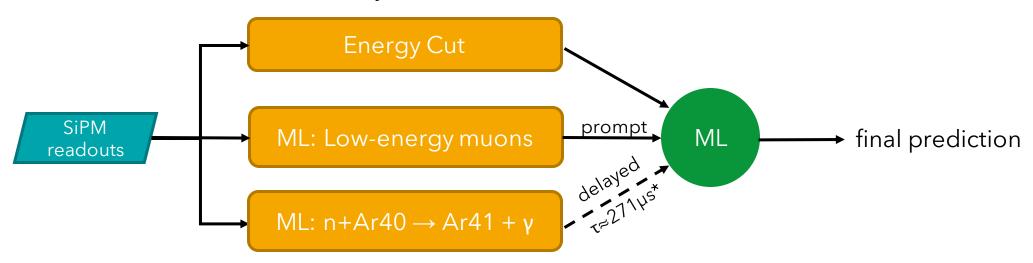
Next Steps

- Detection **Delayed Process** and correlation Prompt/Delayed Classification
- Investigation on sensitivity of trained models w.r.t. realistic experimental conditions
- Consideration on **inference time** to move to an online realm (*trigger implementation*)

BACKUP SLIDES

Proposed Strategy

- Multidimensional problem: depositions in time and space
- Investigate on the use of ML to synthesize an automated classifier
- Leveraging the spatial distribution of depositions to identify specific topologies
 - Low-energy muons (partially cross the sensitive region)
 - Undetected muons but delayed Ar41 de-excitation: $n+Ar40 \rightarrow Ar41 + \gamma$



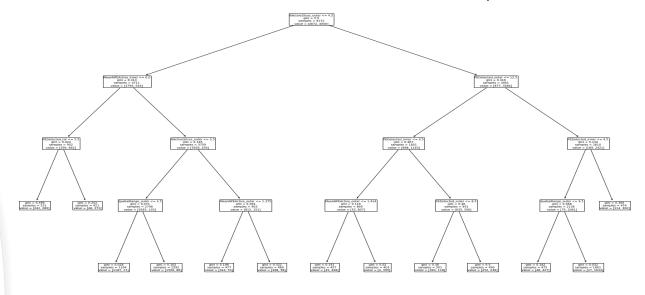
^{*}Wiesinger, et al. "Virtual depth by active background suppression: revisiting the cosmic muon induced background of Gerda Phase II."

Prompt Classification - DTree

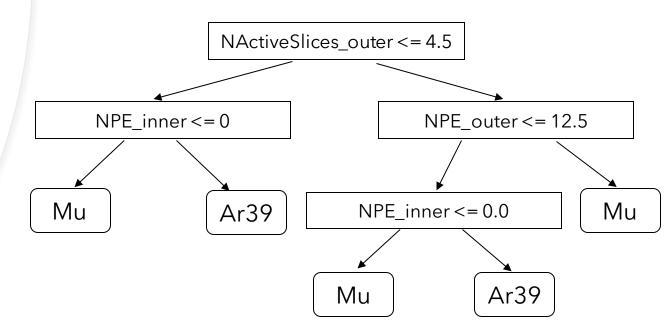
Feature Extraction:

- NPE, MeanNPE, ...
- Nr Active Slices, Spatial Range, Spatial Var, ...
- **Feature Selection**: Remove too-high correlated features
- **Training**: Tune a model on selected features (*top figure*)
- Refinement: Remove redundant features and prune DTree (bottom figure)

Trained Model: 10 Features, Max Depth 4



Pruned Model: 2 Features, Max Depth 3



DTree - Zoom Trained Model

