



ML4NP: Meeting 14.07.2020



Summary

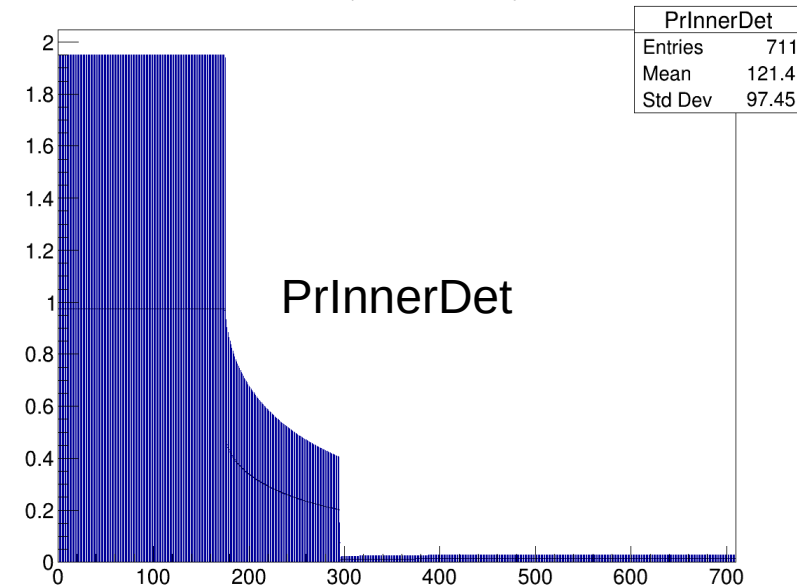
- Spatial Map:
- ML Part I: Rejection of single Ar39 decay
- ML Part II: Classification Muons vs Ar39 Pileups

Spatial Map (last week)

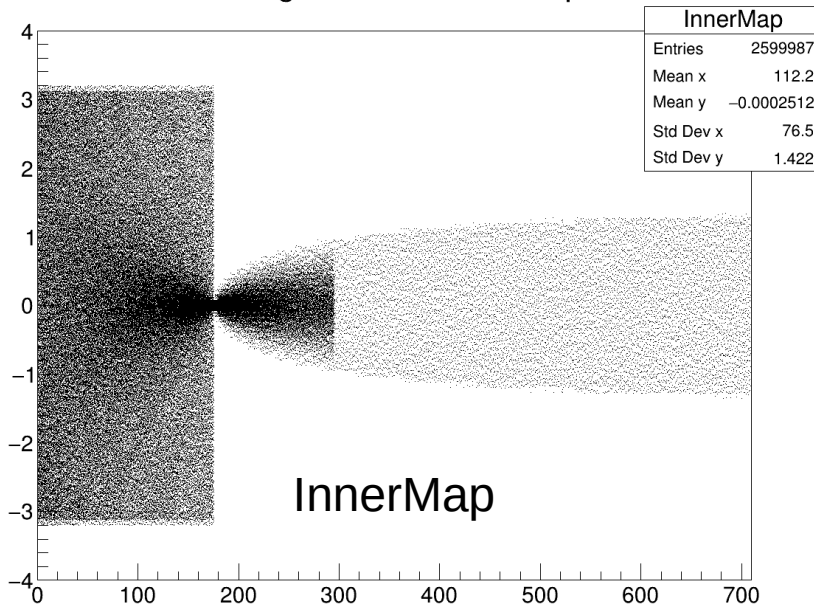
- Production of 3 maps:

- 1) PrInnerDet: given NPE , it computes NPE_{in} that hit the InnerShroud and NPE_{out} that hit the OuterShroud
- 2) Spread NPE_{in} in InnerShroud wt InnerMap
- 3) Spread NPE_{out} in OuterShroud wt OuterMap

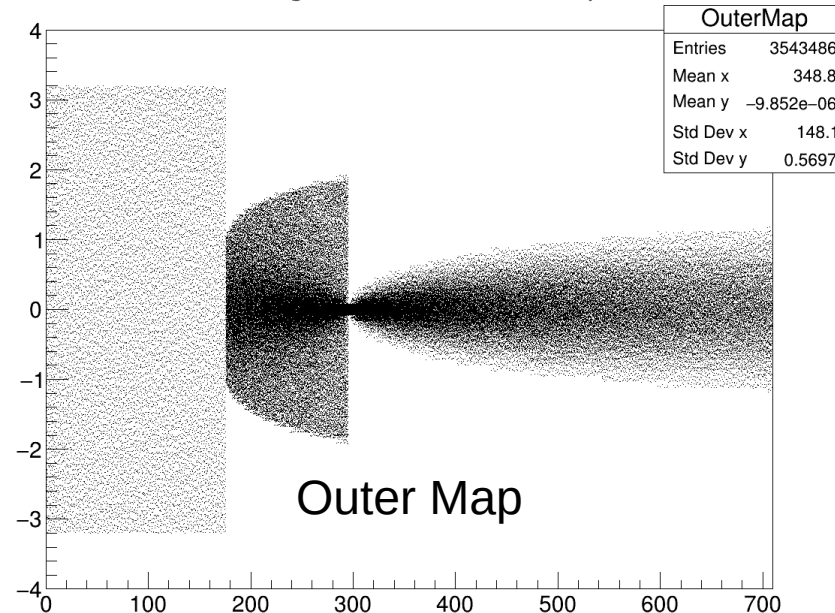
Pr ~ Fract. Inner/(Inner+Outer) Detections



R-Angle Inner Shroud Map

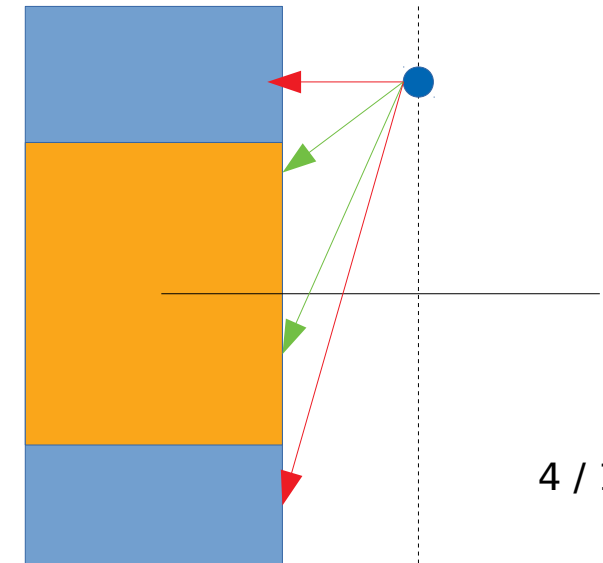
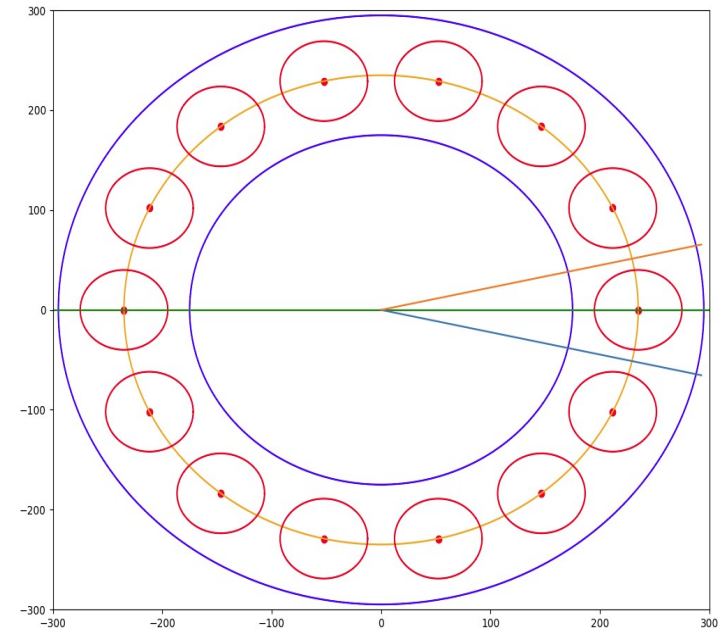


R-Angle Outer Shroud Map



Spatial Map (updates)

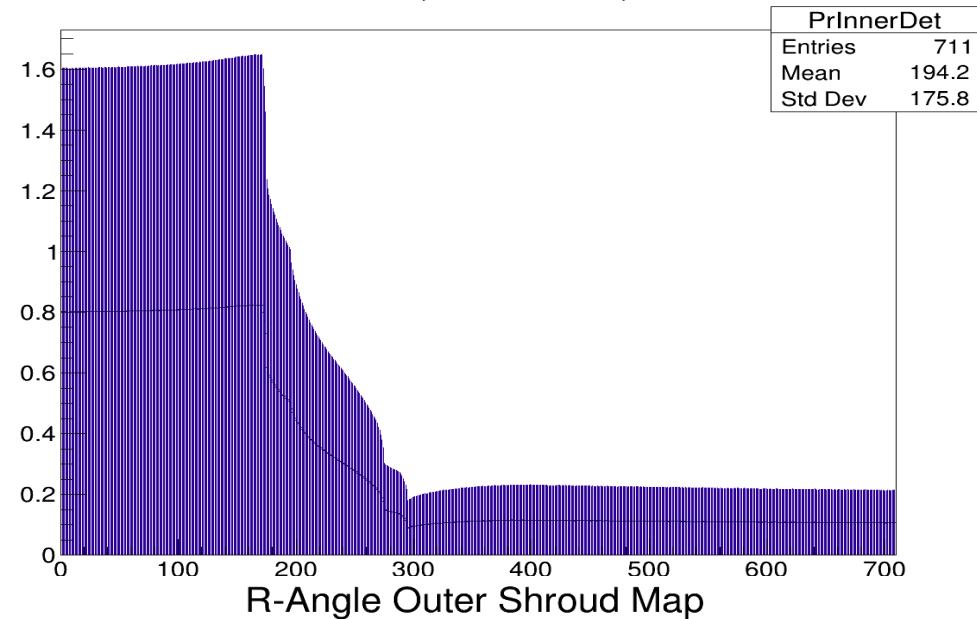
- Inclusion of Ge crystals
(14 strings in geometry = 14 circles in 2D)
- We always sampled in point along a line, extend with sampling in a range ± 0.22 rad.
- The chance to hit Ge crystals depends on:
 - Starting z
 - OP emission angle
- By sampling z , OP phi and OP theta, we compute where the OP trajectory cross the Ge surface (cylinder $23.5+40$ mm).
 - If the crossing point has z in ± 445 mm, enable Ge (*green arrows*).
 - Otherwise, disable Ge (*red arrows*).
- The OP trajectories ends when they go outside the roi. We cut a 2D trajectory based on where the 3D trajectory cross ± 845 mm.



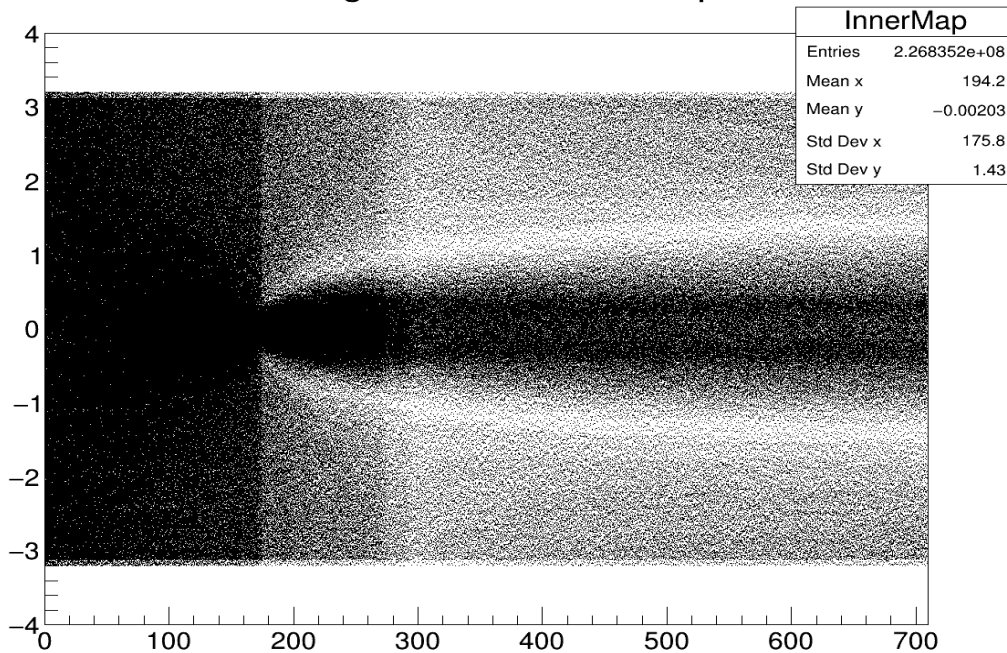
Spatial Map (updates)

- The new maps include:
 - sampling of shifting angle and Z,
 - sampling of OP orientations,
 - enabling/disabling Ge crystals,
 - early cutting of OP trajectories.

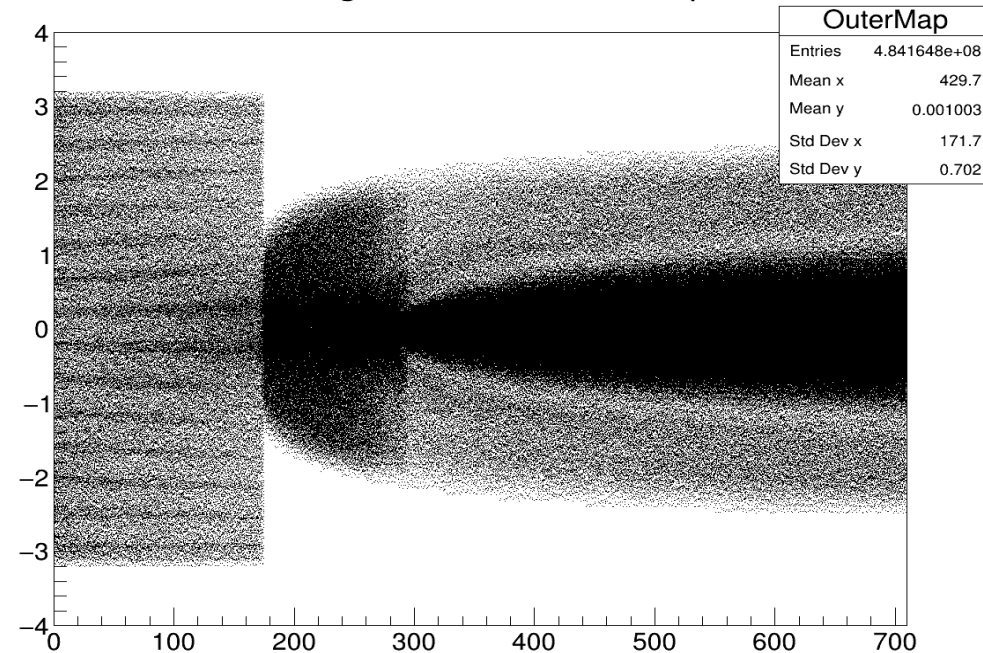
$Pr \sim \text{Fract. Inner}/(\text{Inner}+\text{Outer}) \text{ Detections}$



R-Angle Inner Shroud Map



R-Angle Outer Shroud Map



ML Part 1: Rejection of Ar39

- **Context:**
 - Very low-energy events: detections up to 60 PE
 - Level 0 Tagging: specialized in 1 Ar39 to suppress its high-activity
- **Requirements:**
 - 1) Reject single Ar39 decays as much as possible (high TNR),
 - 2) Save very low-energy muons ($PE \leq 60$), if possible.

ML Part 1: Rejection of Ar39

- **Available Data:** 5 834 261 Ar39, 26 881 Mu (*old data + new 1M muons by CJ*)
- **Preliminary cut at 5 PE** to discard most of the single Ar39 decays (*avg 3 PE*):
 - 5 110 Muons, 1 474 688 Ar39
- **Training Data** (*balanced*): 4 500 Muons, 4672 Ar39s (undersampling)
- **Test Data** (*unbalanced*): 610 Muons, 1M Ar39s
- Trained model:
 - Evaluation on Test Data:
 - Accuracy: 0.880, Purity: 0.004, Efficiency: 0.852, F1: 0.009
 - Conf. Matrix:
TN: 879742, FP: 120258,
FN: 90, TP: 520
 - **TPR: 85.25%, FPR: 12.03%** (*without considering preliminary cut*)

From model to code...

```
PEDetected_inner, PEDetected_outer, NActiveSlices_outer = 0, 1, 2    # to call the position
target_ar, target_mu = 0, 1
def l0_dtree_func(row):
    # `row` is a line in the format PEDetected_inner, PEDetected_outer, NActiveSlices_outer
    row = row.to_numpy()
    if (row[PEDetected_inner]+row[PEDetected_outer]<5):
        return target_ar
    if(row[NActiveSlices_outer] <= 4.5):    #NActiveSlice_outer <= 4.5
        if(row[PEDetected_inner] <= 0.5):
            return target_mu
    elif(row[PEDetected_outer] > 12.5):    #NActiveSlice_outer > 4.5
        return target_mu
    elif(row[PEDetected_inner] <= 0):    #NActiveSlice_outer > 4.5 & PEDetected_outer<=1
        return target_mu
    return target_ar
```


ML Part 1: Rejection of Ar39

- Passing the various Ar39 classes through this selection strategy.

```
[Info] Dataset: 1 Ar39 Pileup, FPR: 0.028126871845056918
[Info] Dataset: 2 Ar39 Pileup, FPR: 0.21759204222353887
[Info] Dataset: 3 Ar39 Pileup, FPR: 0.39855888088584174
[Info] Dataset: 4 Ar39 Pileup, FPR: 0.5169477984980941
[Info] Dataset: 5 Ar39 Pileup, FPR: 0.6432688146817842
[Info] Dataset: 6 Ar39 Pileup, FPR: 0.7805673699507727
[Info] Dataset: 7 Ar39 Pileup, FPR: 0.8912115042136624
```

ML Part 2: Mu vs Ar39 Pileups

- **Context:**
 - Low-energy events: detections up to 115 PE
 - Level 1 Tagging: wide classification 1, 2, ..., 7 Ar39 decays vs Muons
- **Requirements:**
 - Prompt tagging of muons, suppression of Ar39 (single or pileup)
 - Good metrics: Efficiency/Purity
- First attempt:
 - **CNN model:** slower inference but no cost for feature creation (*at least on my pc, e.g. 701000 instances in 16 seconds*)
 - The previous Dtree has faster inference (*e.g. 901235 instances in 8 seconds*) but requires feature creation