



Brandeis  
UNIVERSITY

# Update on Particle ID

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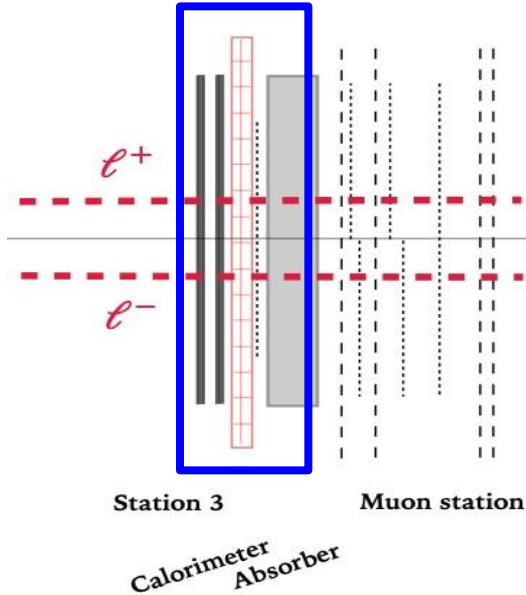
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W.P. McCormack, P.C. Harris (MIT)

Oct. 2023



# INDEX

- Introduction about particle ID
  - Current version of ID
  - New Clustering algorithm
  - Optimized analysis package
- 
- Total # of page: 27
  - Estimate time: 20mins

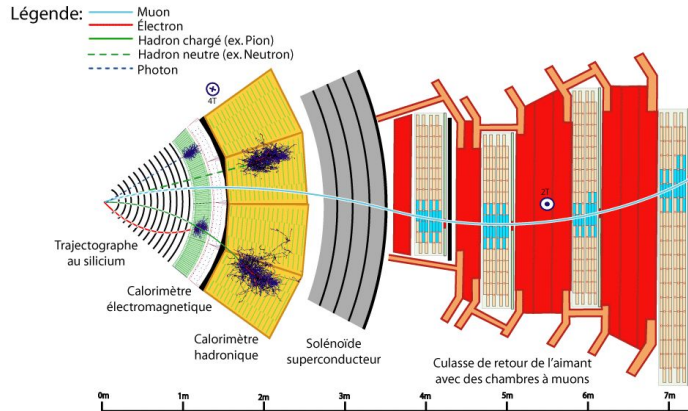


# Introduction to ID

- Short introduction & goal
- Data pattern for different particles
- Historical versions

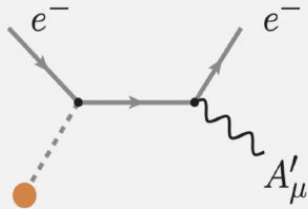
**Particle ID** aims to classify different particles in each event, where a particle is reconstructed using multiple detector components. This information helps us to form a complete picture of what happened in events.

The ultimate goal for our work is to build a particle ID like tagging system of particle flows at CMS. That we use **tracking chambers**, **calorimeter**, and **hodoscope** information to reconstruct trajectories and tag particles in each single event.

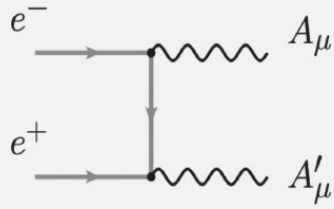


A considerable part of our work is exploratory, and highly customized for DarkQuest experiment setup.

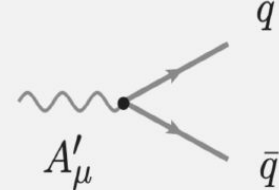
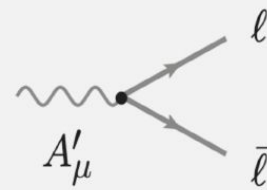
# Dark Photon Production and decays



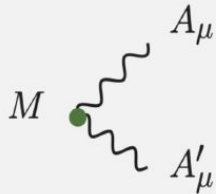
*Bremsstrahlung*



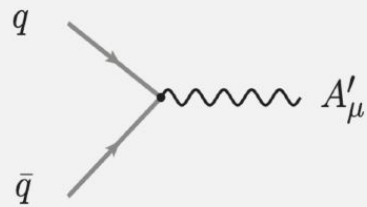
*Annihilation*



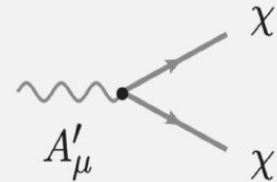
*visible*



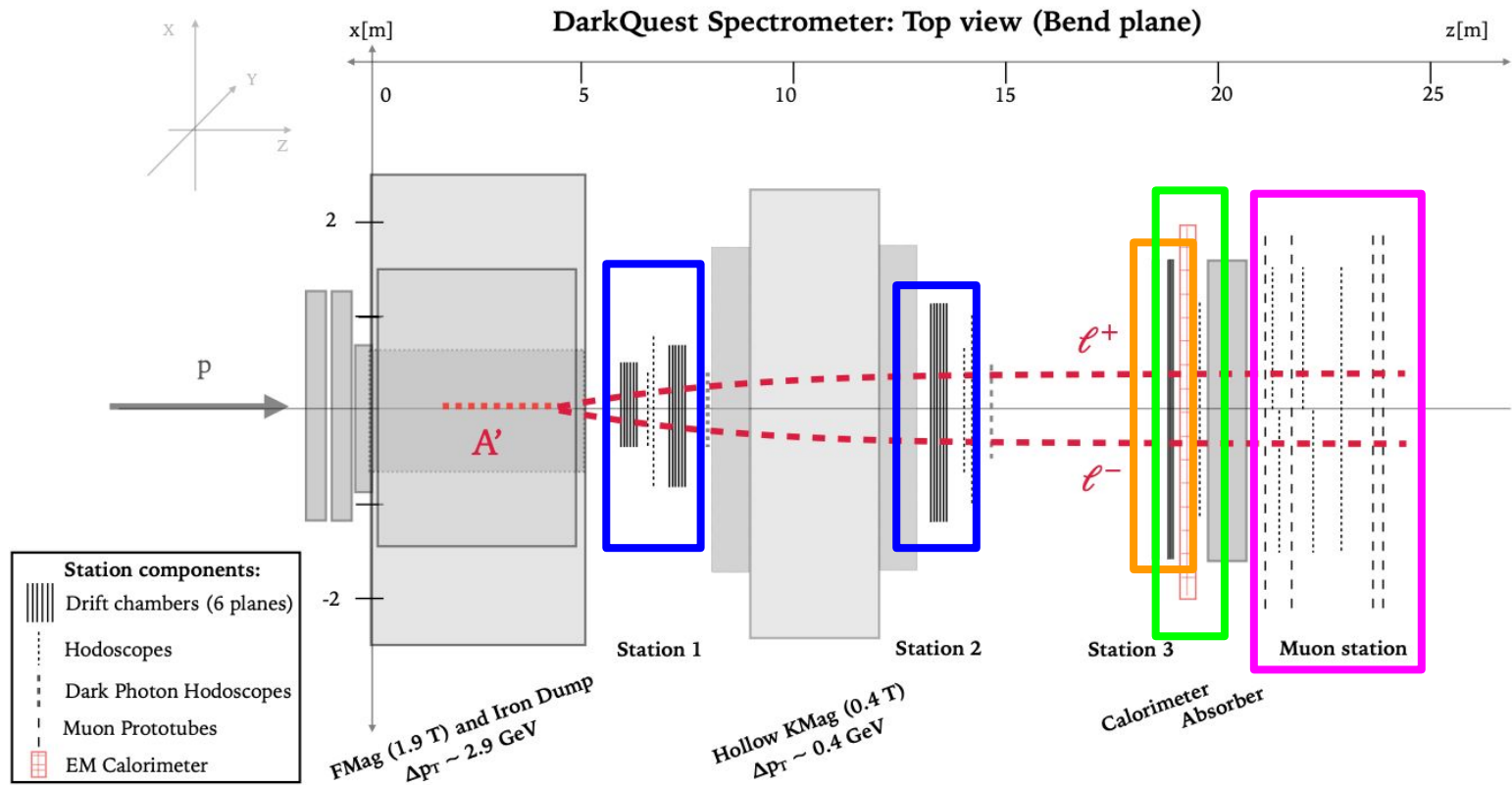
*Meson decay*



*Drell-Yan*



*invisible*



Full track(st1+st2+st3)

st3\_stracklet

EMCal

Hodoscope

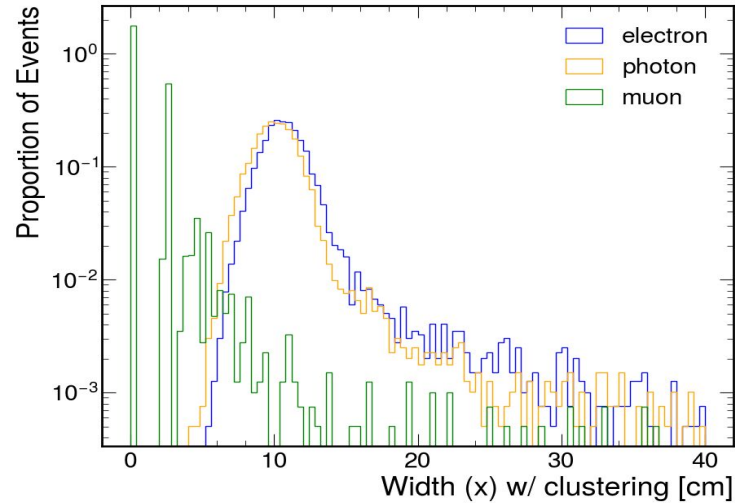
[E/p]

[x, y, z, px, py, pz]

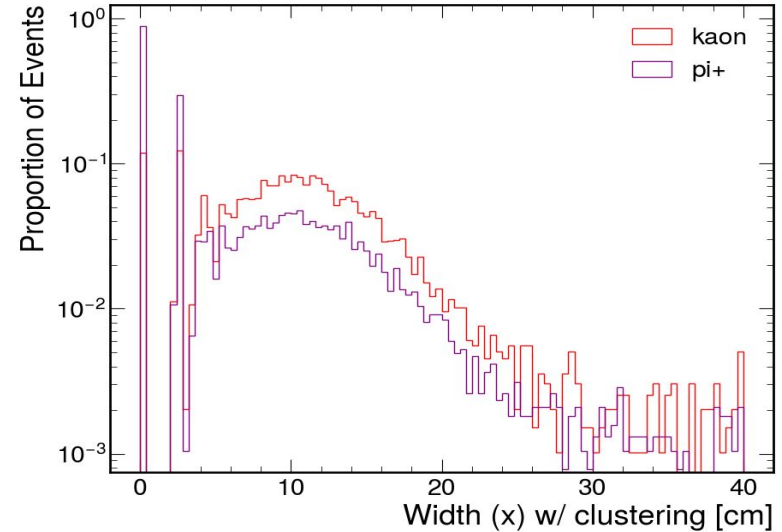
[widthx, widthy, Wew\_x, Wew\_y, seed\_x, seed\_y]

[det41\_y, det42\_y, det43\_y, det44\_y, det45\_x, det46\_x]

# Quantifying cluster shapes



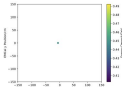
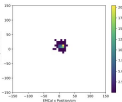
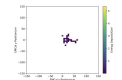
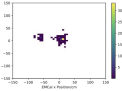
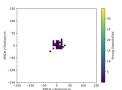
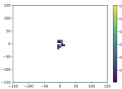
Full track(st1+st2+st3) , st3\_stracklet, Hodoscope. Only has a single value for each particle.



To quantitatively study the shape of cluster on Emcal, we plot the distribution of width(x and y), Energy Weighted width of particles on EMCal. As in DQ

Along with track extrapolation points on EMCal, Energy over momentum, we implemented the first version of ParticleID.

# Spreadsheet for particle characteristics

Types	Full track (E/P)	st3_tracklet	EMCal Cluster size	Hodoscope
Muon	Yes, ~0	Yes	Few Points 	Yes, through absorber
Electron/Positron	Yes, ~1	Yes	Large 	No
Pi+/Pi-	Yes, (0,1)	Yes	Middle 	Some
Pi0	No	No	XLarge 	No
Photon	No	No	Large 	No
Klong	No	No	Middle 	No



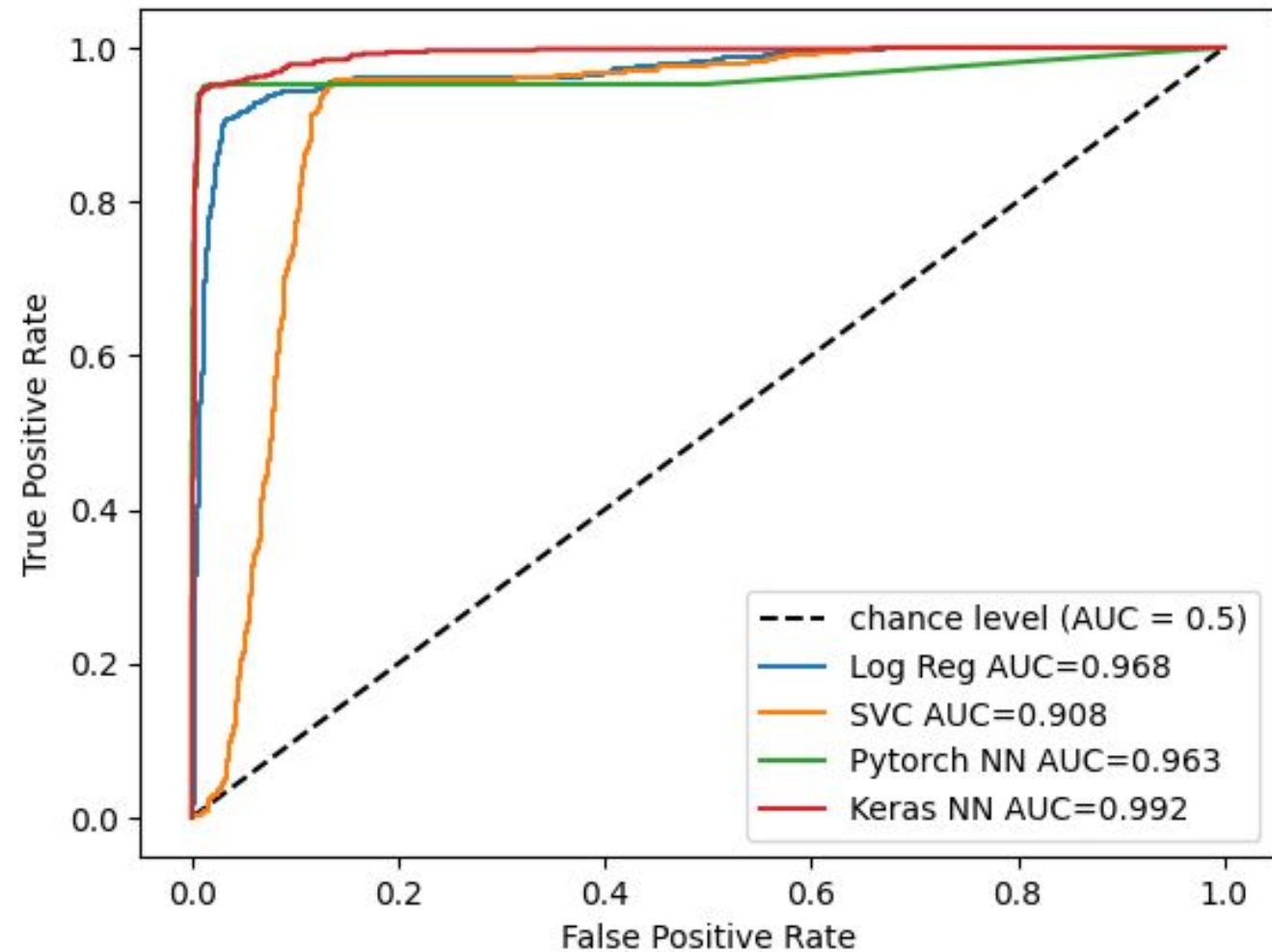
# Cut-Based ID for Electron

	Pass Rate
muon	0
electron	59.50%
positron	55.70%
photon	0.40%
pi+	3.40%
pi-	3.80%
klong	0
pi0	0.20%

Inputs:  
[widx, widy, Wew\_x, Wew\_y, Dist\_x,  
Dist\_y, E/P]

Well, can we do a DNN based particle ID to improve the efficiency?

OvR ROC curve for electron

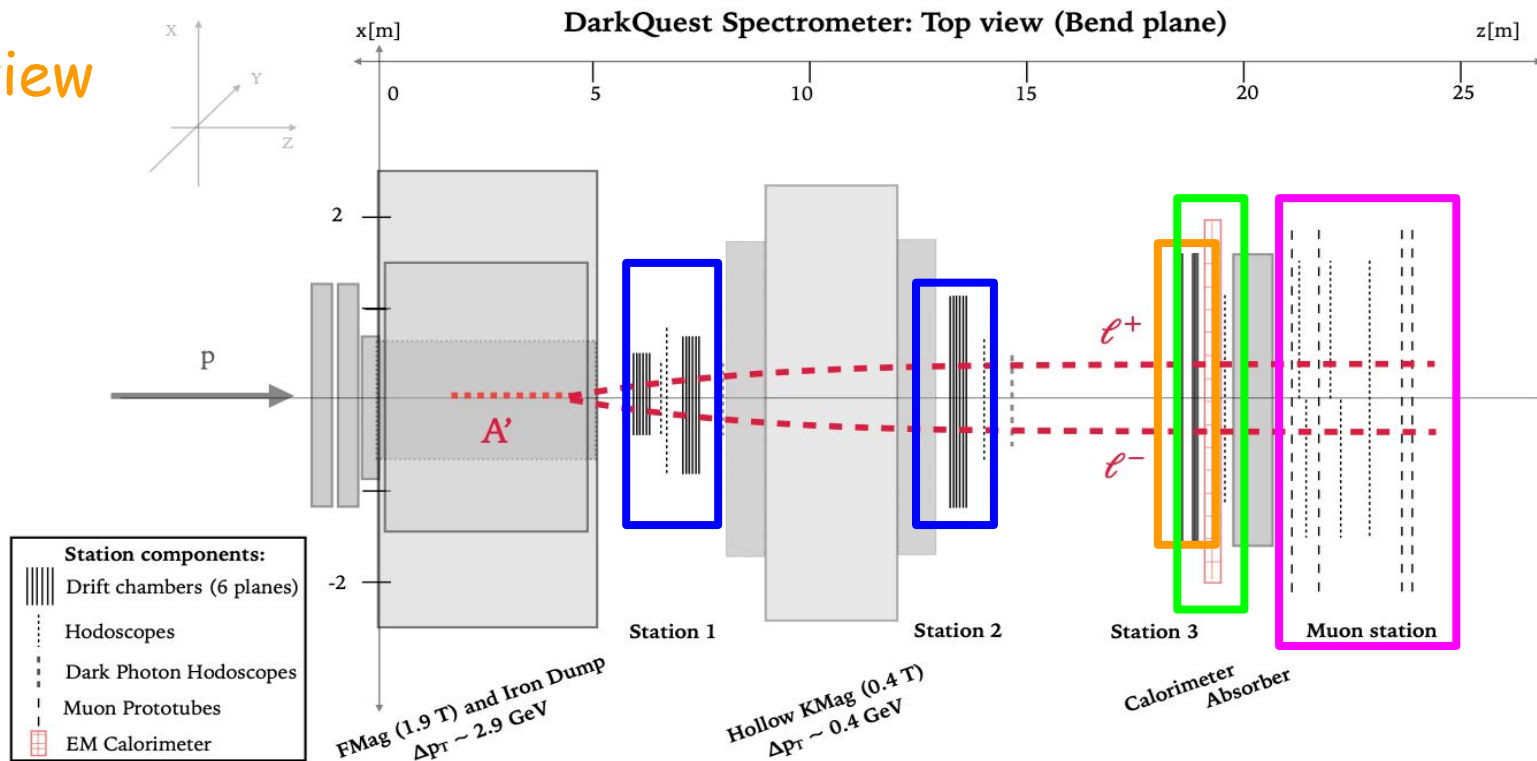


This NN-based ID has the same input as cut-based one, looks good?

The analysis is only on leading cluster of each single particle guns evts.

But we are looking for dark photon into dimuon and dielectron, right?

# Review



Full track(st1+st2+st3)

St3\_stracklet

EMCal

Hodoscope

[E/p], effectively distinguish between muon and electron

Position and direction to extrapolate trajectory

\*Currently use # of seeds to determine how to reconstruct

Finally will depends on #of trkls and EMCal energy seeds

Efficiently get muon and  $\pi^\pm$  separated.

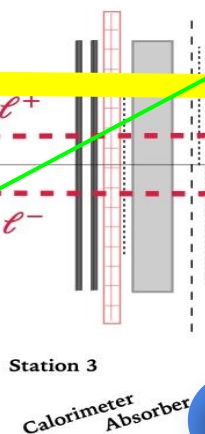
# Reconstruction Logic

1 Track finding & ECal clustering

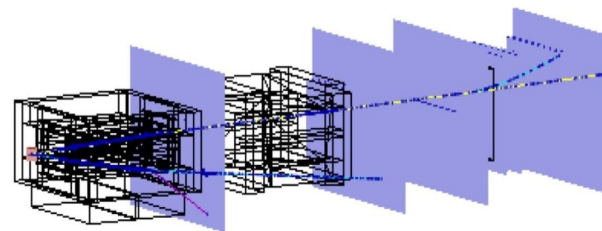
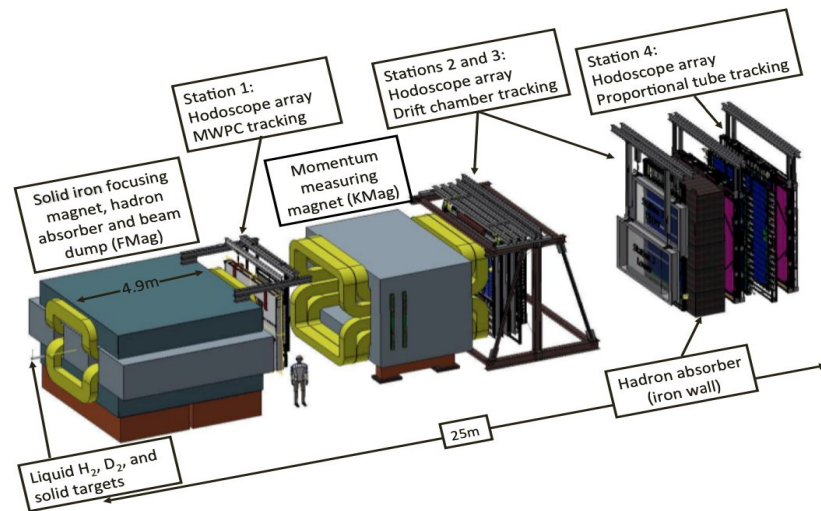
2 Clusters match track & hodoscope hits

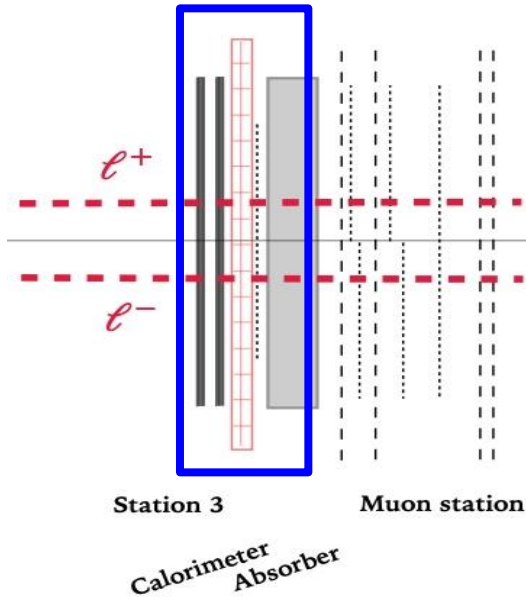
3 \*post ana for "missing muon"

4 Run tagger on "particles"



$\mu^+ -$

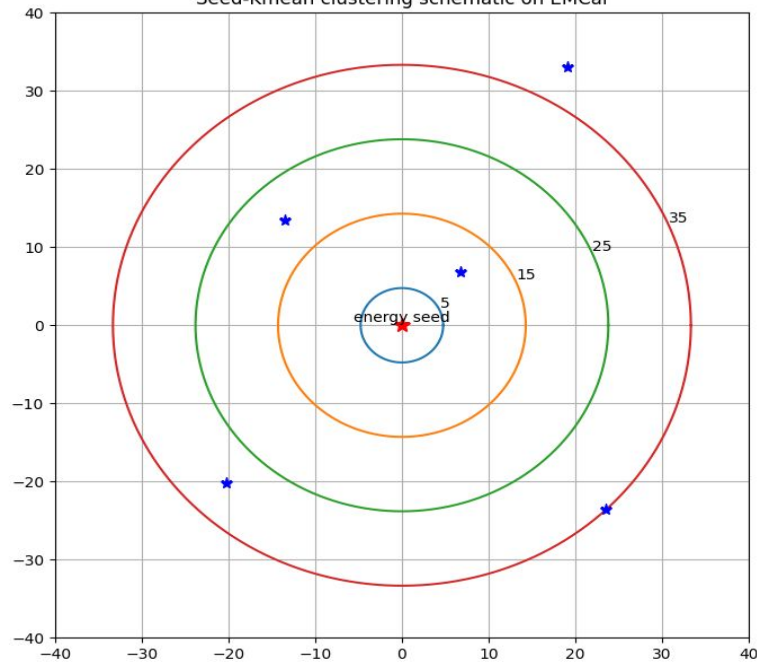




## New Clustering & Efficiency

- $A' \rightarrow \mu\mu$
- $A' \rightarrow ee$
- Single particle guns
- Diphoton samples

Seed-Kmean clustering schematic on EMCal



**Step3.** Run Kmean once, find if some cluster has been missed (mainly for muon). \*exceed tolerance.& (post-ana)

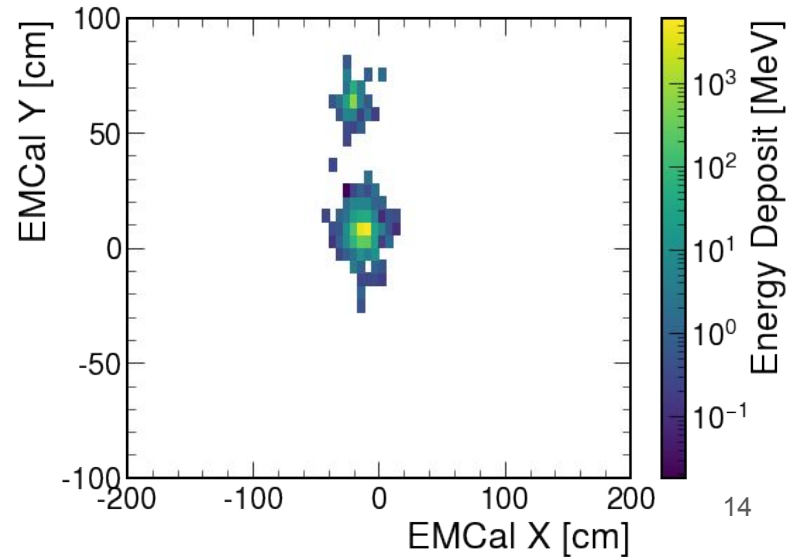
**Step4.** Once get seeds all set, tagging each hits to clusters.

**Step1.** Set a energy threshold,  $e_{min}$ . Collect all hits with eng above this level.

$[x_i, y_i, eng_i]$

**Step2.** Loop through the hits, within distance threshold, no hits have higher energy.

\*currently set  $e_{min}=0.2\text{GeV}$ , radius=50cm. Low energy threshold make sure hard to miss a particle.

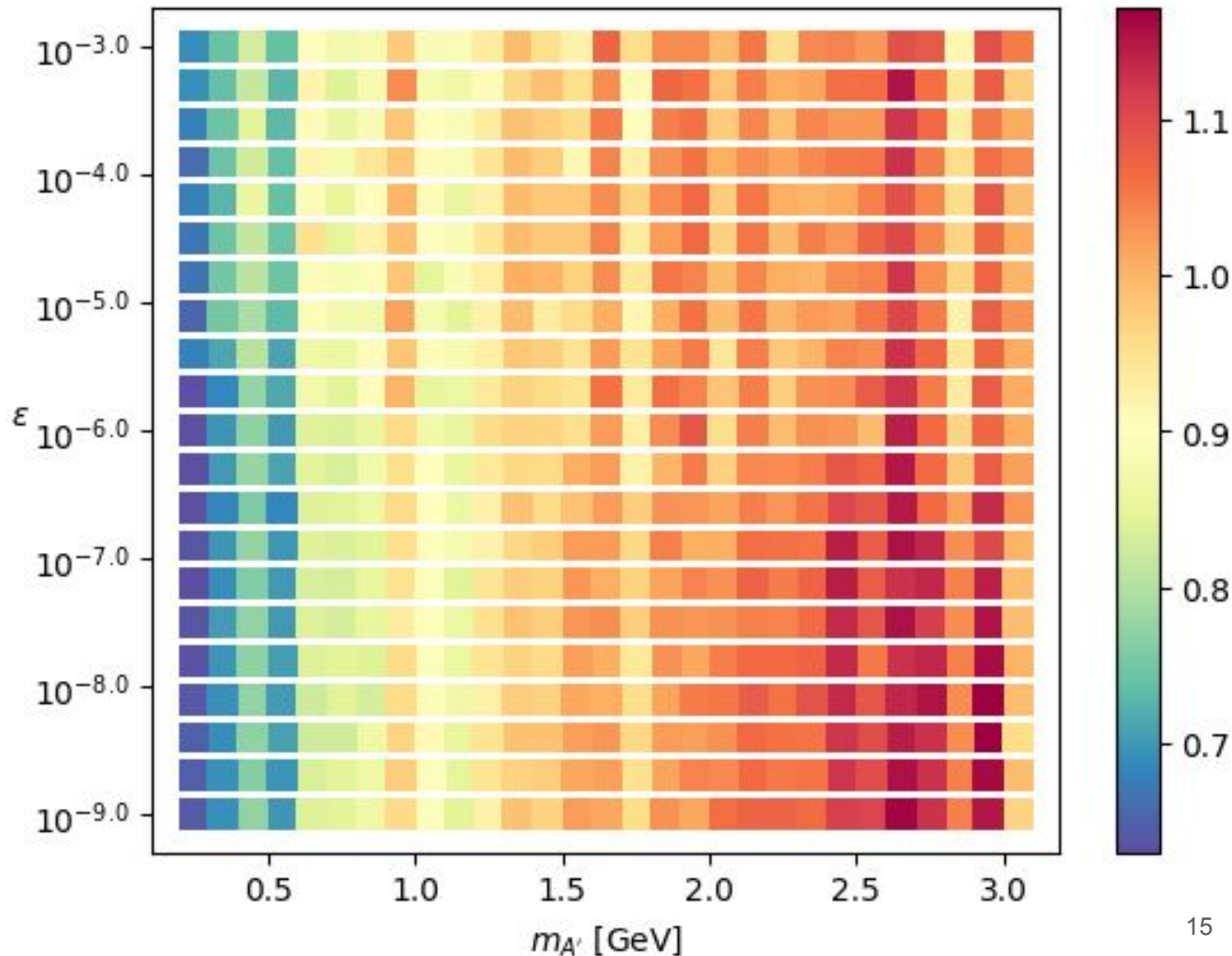


$A' \rightarrow \mu\mu$

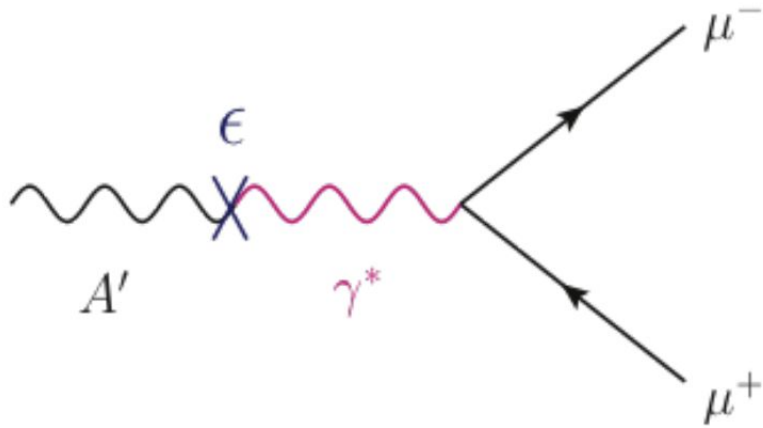
Seed tagging efficiency

All 609 possible  
mass and coupling  
combination.

$$\text{eff\_st} = \left( \frac{\sum \# \text{ seeds}}{\sum \# \text{ trkls}} \right)$$



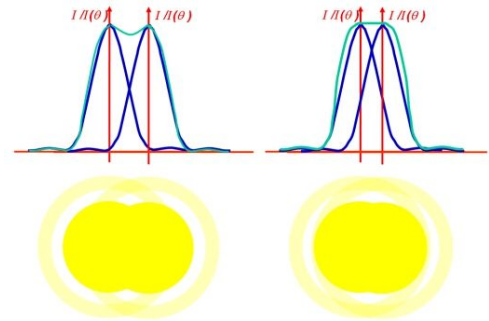
# What happened to low mass $A'$ ?



\*From Ming

EMCal  
cell

Like Angular  
resolution

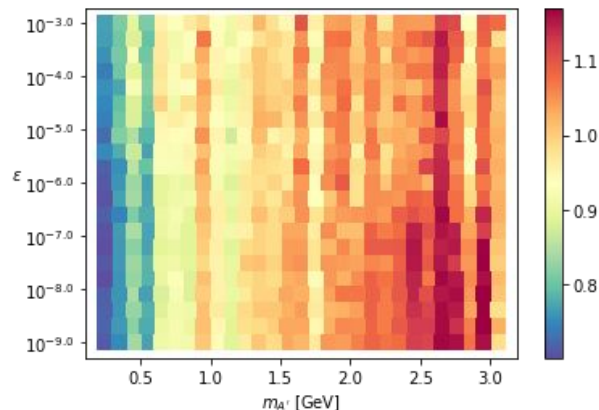


\*Diphoton has the same problem



# Verification of Guess

- Even lower the energy threshold and radius to 0.1, 10. Still doesn't give visible difference.
- Conclusion after manually checking dark blue samples: there are higher density of events has only one hit on EMCAL

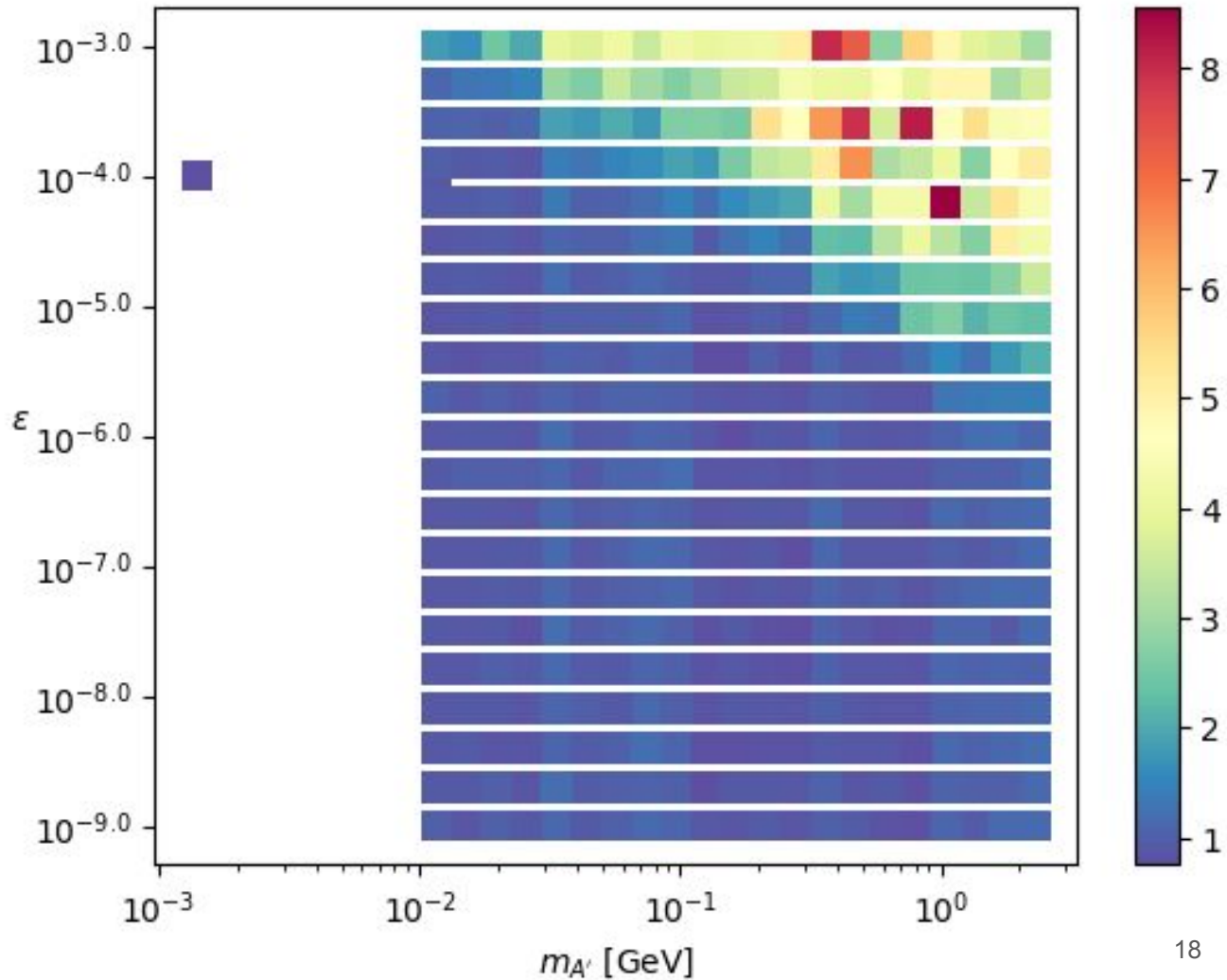
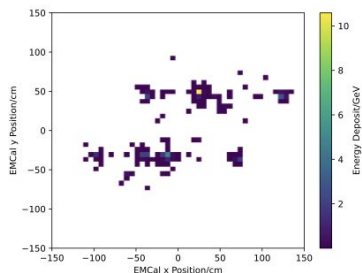


# $A' \rightarrow ee$

Seed tagging efficiency

All 443 possible  
mass and coupling  
combinations.

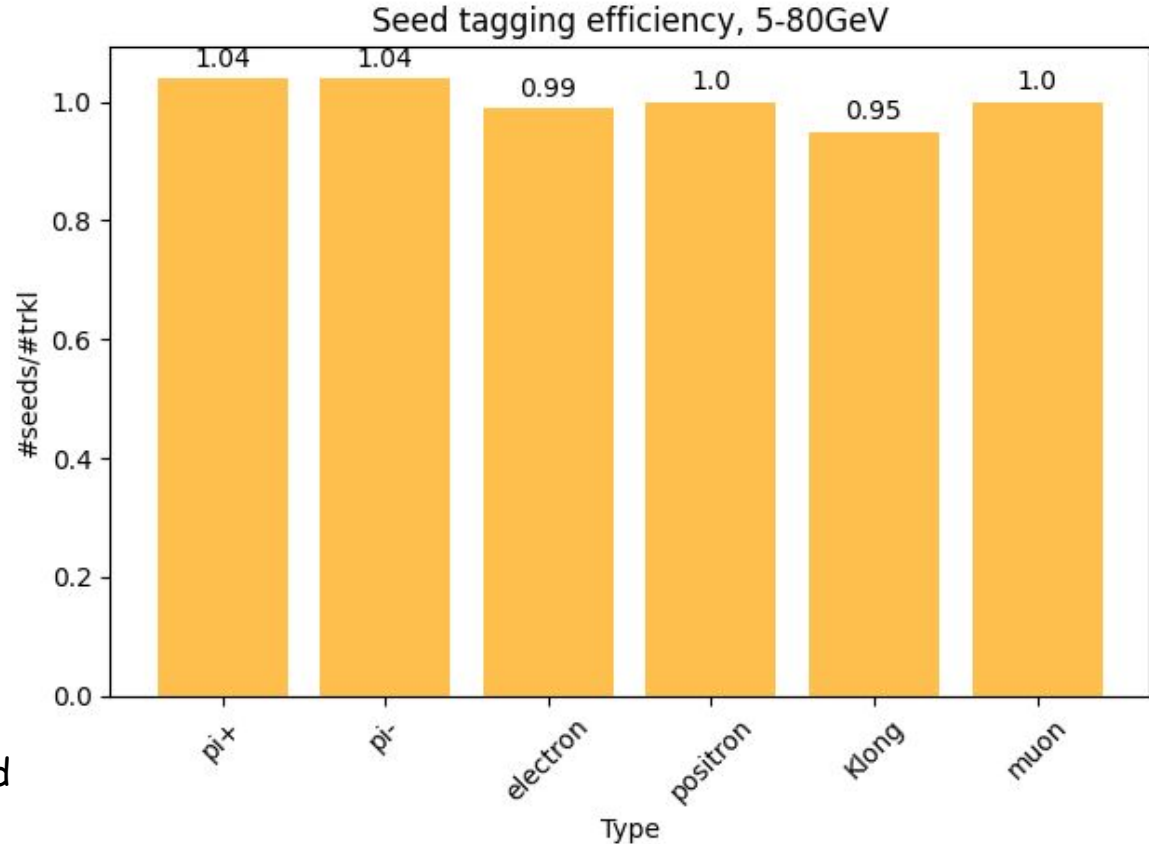
Dark red



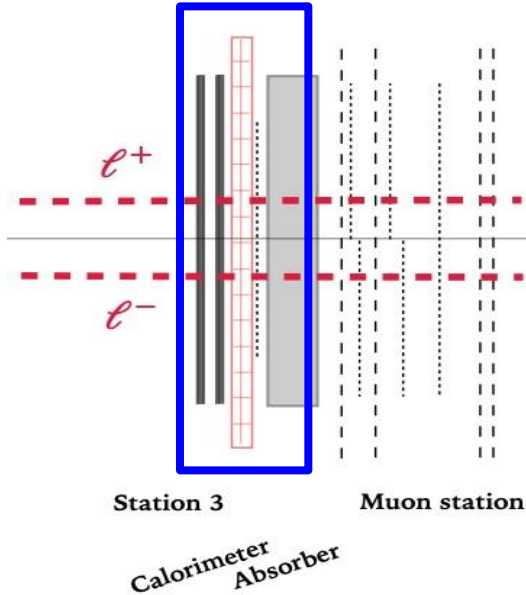
# Seed tagging efficiency of single particle gun

1=100%

$$\text{eff\_st} = \left( \frac{\sum \# \text{ seeds}}{\sum \# \text{ trkls}} \right)$$



\*For Klong, ratio is #evts w/ seed / evts w/ valid EMCAL record



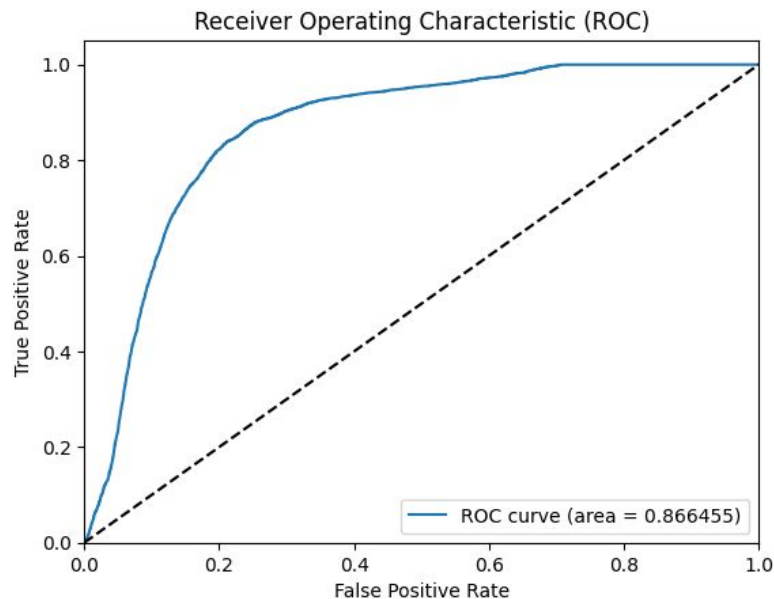
## New Particle ID

- 2 versions w/ and w/o track
- For both muon and electron
- Tagging efficiency vs  $P_z$

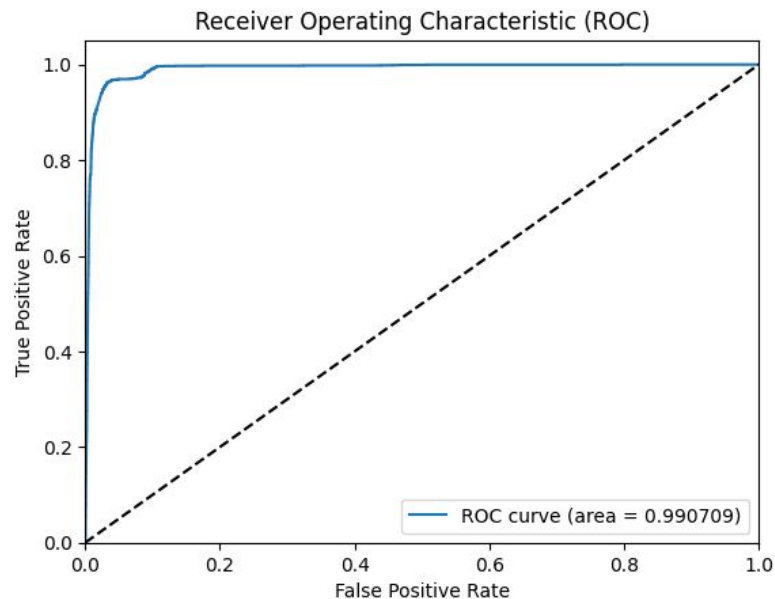
- With the reconstruction logic mentioned above, we can connect hits on each station to a consistent particle trajectory.
- With information array for each particle as NN input. Save into CSV  
`[wid_x, wid_y, wew_x, wew_y, seed_x, seed_y, trkl_x, trkl_y, trkl_z, trkl_px, trkl_py, trkl_pz, E/p, h4_41, h4_42, h4_43, h4_44, h4_45, h4_46]`
- Preparing with my package dwong, mention later.



## Result: ROC curve for ID w/o full track(E/p)



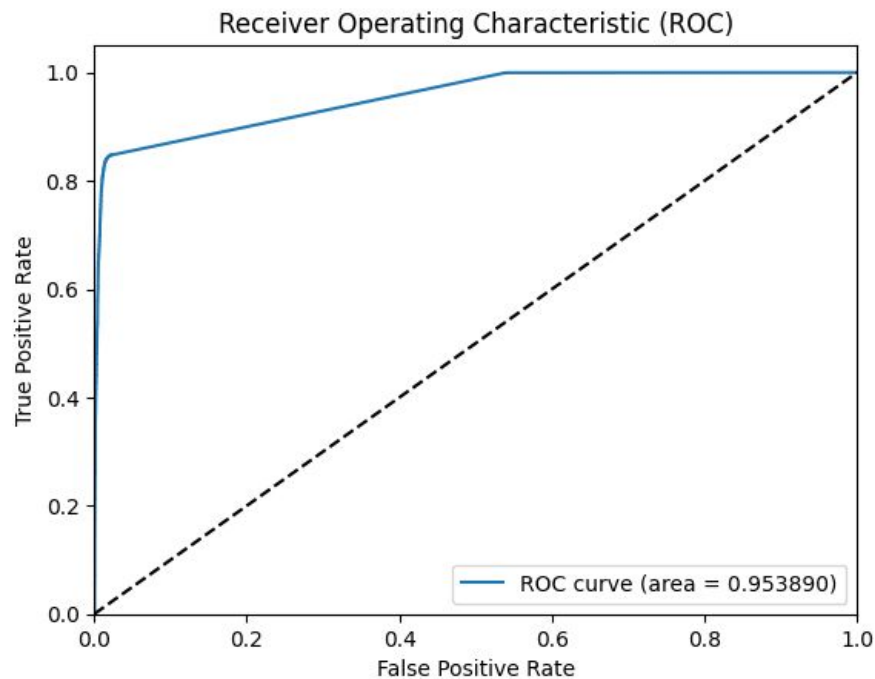
Electron ID



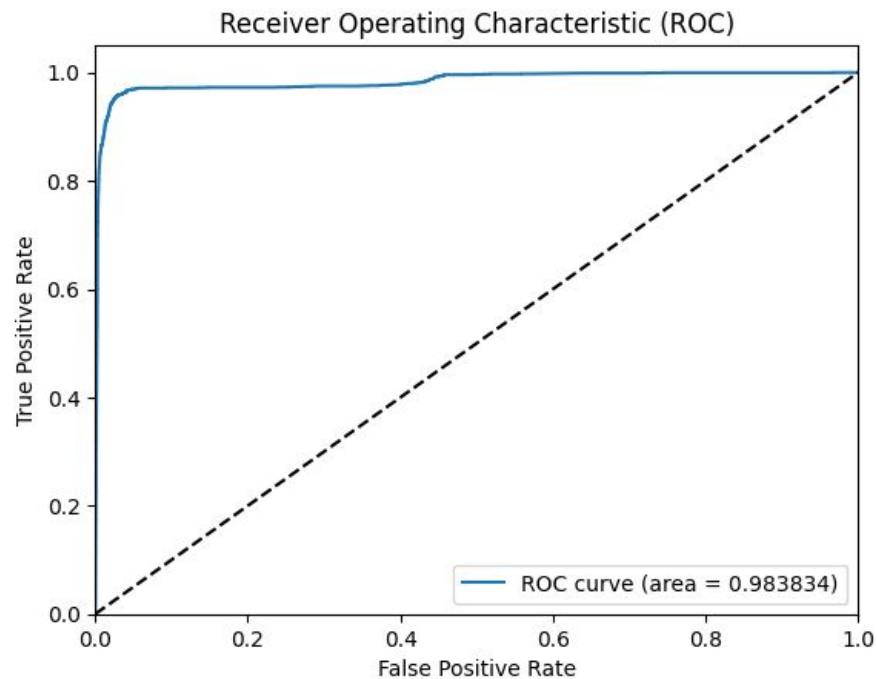
Muon ID

Input: [wid\_x, wid\_y, wew\_x, wew\_y, seed\_x, seed\_y, trkl\_x, trkl\_y, trkl\_z, trkl\_px, trkl\_py, trkl\_pz, h4\_41, h4\_42, h4\_43, h4\_44, h4\_45, h4\_46]

## Result: ROC curve for ID w/ full track(E/p)



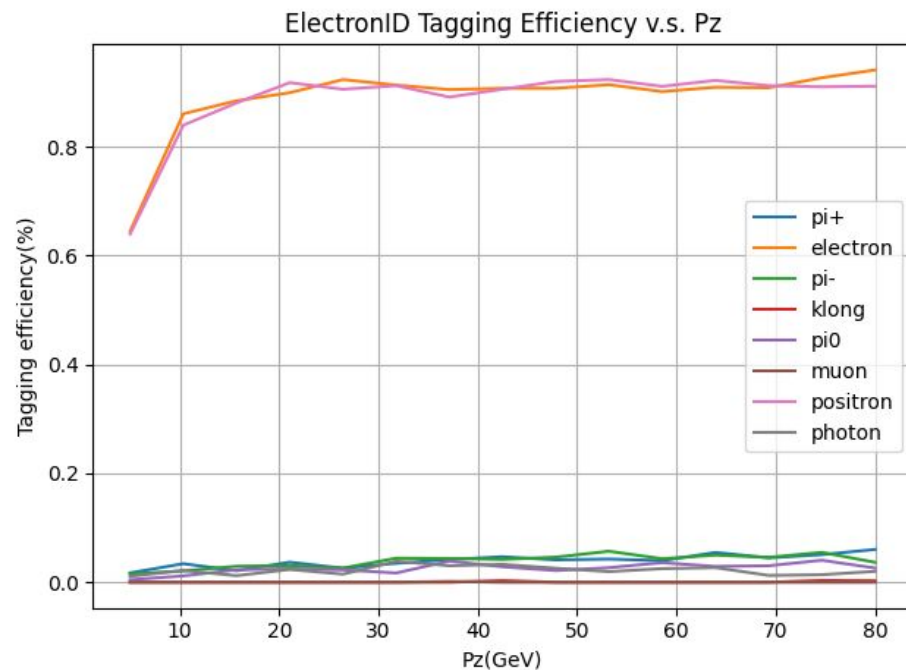
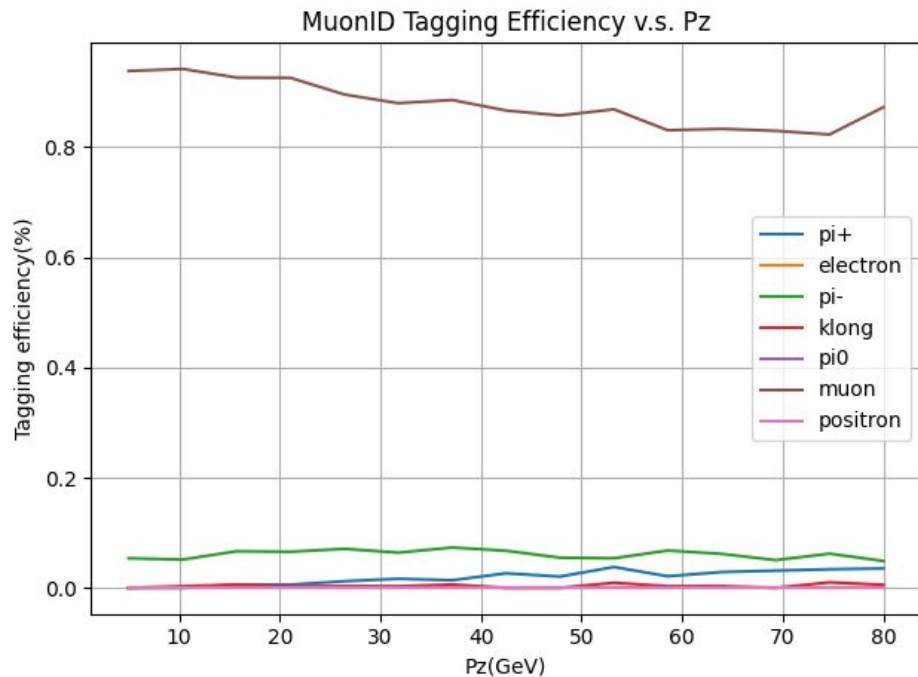
Electron ID



Muon ID

Input: [wid\_x, wid\_y, wew\_x, wew\_y, seed\_x, seed\_y, trkl\_x, trkl\_y, trkl\_z, trkl\_px, trkl\_py, trkl\_pz, E/p, h4\_41, h4\_42, h4\_43, h4\_44, h4\_45, h4\_46]


# Result: Tagging efficiency vs $P_z$



Why range 5-80GeV?

**DarkQuest: A dark sector upgrade to SpinQuest at the 120 GeV Fermilab Main Injector**





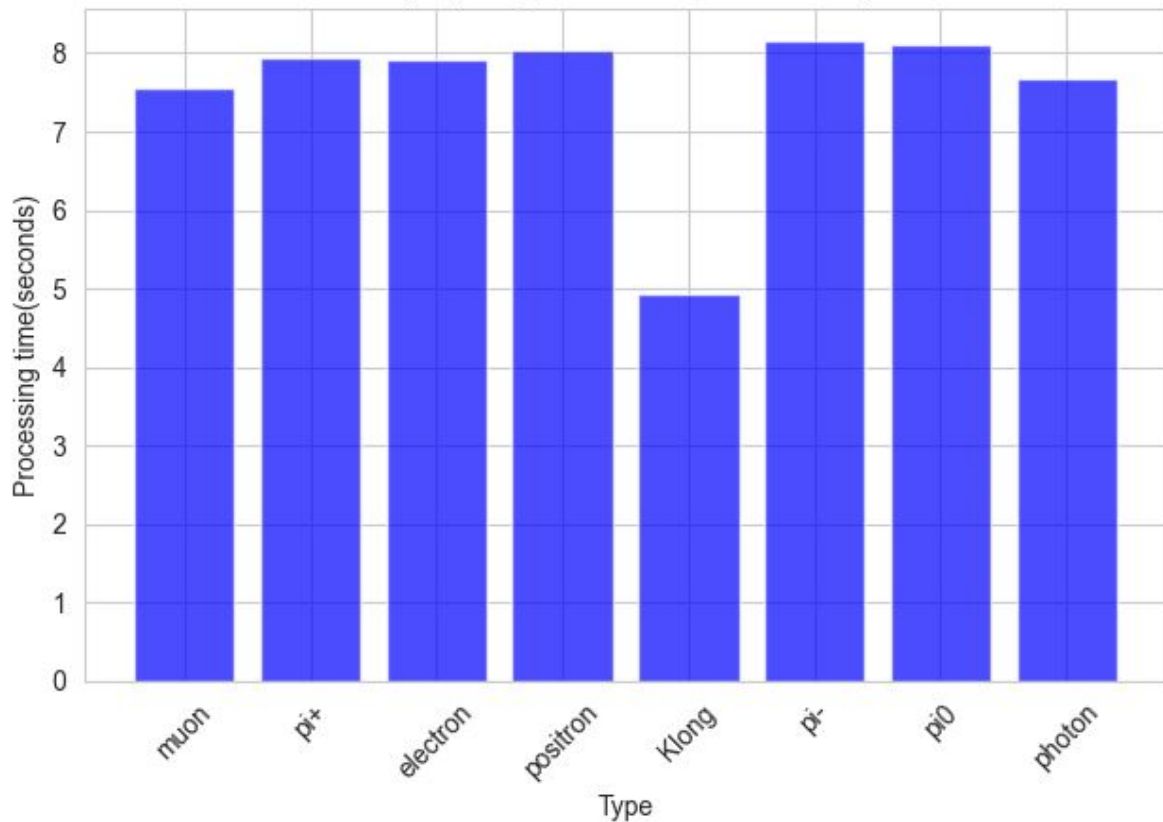
```
def getData(fname="", procName="Events"):
    file = uproot.open(fname)
    dq_dict = file[procName].arrays(library="np")
    dq_events = {
        "Hits":{
            "detID": dq_dict["hit_detID"],
            "edep": dq_dict["hit_edep"],
            "elmID": dq_dict["hit_elmID"],
            "hit_pos": dq_dict["hit_pos"]
        },
        "track":{
            "x": dq_dict["track_x_CAL"],
            "y": dq_dict["track_y_CAL"],
            "ID": dq_dict["eventID"],
            "pz": dq_dict["track_pz_st1"]
        },
        "st23": {
            "ntrack23": dq_dict["n_st23tracklets"],
            "px": dq_dict["st23tracklet_px_st3"],
            "py": dq_dict["st23tracklet_py_st3"],
            "pz": dq_dict["st23tracklet_pz_st3"],
            "x": dq_dict["st23tracklet_x_st3"],
            "y": dq_dict["st23tracklet_y_st3"],
            "z": dq_dict["st23tracklet_z_st3"],
            "Cal_x": dq_dict["st23tracklet_x_CAL"],
            "Cal_y": dq_dict["st23tracklet_y_CAL"]
        }
    }
    return dq_events
```

## Package dwong!!!

- DQ analysis package dwong released on pip. People could play around.
- Include analysis functions, NN-ID model, training frame, csv saving tools, samples and EMCal plot tools.
- Optimized with numpy and cython.

# *It's all open source!*

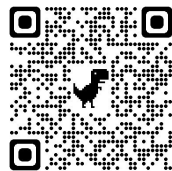
Data preparing time for 10,000evt samples



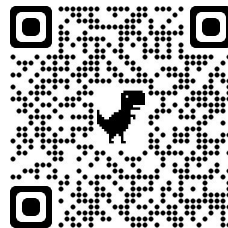
[Repo for analysis code:](#)



[Repo for dwong package:](#)



[Pypi page of dwong](#)





# dwong 0.1



Latest version

```
pip install dwong==0.1
```



Released: less than 20 seconds ago

Dowling's integrated data analysis for DQ experimnt setup. Include data analysis functions, csv saving tools, Particle ID model, DNN model frame for ID training and EMCal plotting tools

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## Project description

### dwong, a package for DarkQuest.

dwong is a comprehensive Python package, created by student Dowling Wong, tailored for data analysis and neural network-based particle identification in the DarkQuest experiment. The aim of this project is to streamline DarkQuest's data analysis process by providing exemplary data-processing functions.

The package mainly contains four modules: dwong, dplot, dcsv and dkeras.

- dwong
  - emcal\_bytuple
  - multi\_clusters
  - h4\_bytuple
  - prepare\_data\_bytuple
- dplot
  - emcal\_evt(x, y, eng)
  - emcal\_pdf(ntuple\_name, filename, path)
- dkeras

Thank You!  
Questions?