

# Improving the gamma-hadron separation for air showers at the IceCube Neutrino Observatory

Machine Learning workshop @ Delaware University

Karlsruher Institut für Technologie (KIT)  
Institut für Astroteilchenphysik (IAP)

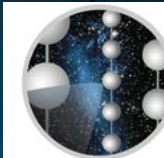
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**ICECUBE**  
SOUTH POLE NEUTRINO OBSERVATORY

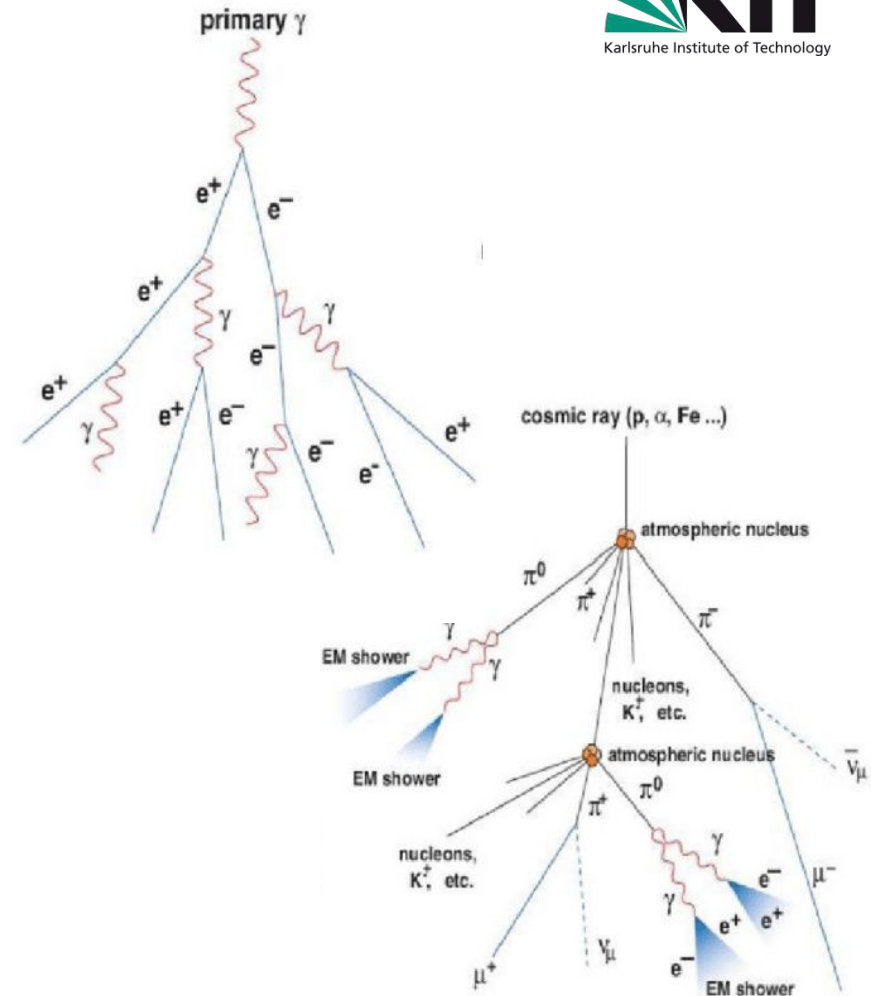
## Previous works

2 PhD students from Delaware University (USA):

- SEARCH FOR PEV GAMMA RAYS WITH THE ICECUBE OBSERVATORY  
Zachary Dean Griffith  
→ Focused on ML techniques for Gamma Hadron discrimination
  
- SEARCH FOR PEV GAMMA RAYS AND ASTROPHYSICAL NEUTRINOS WITH ICETOP AND ICECUBE  
Hershal Pandya  
→ Developed Log Likelihood Ratio parameter

# Gamma hadron separation

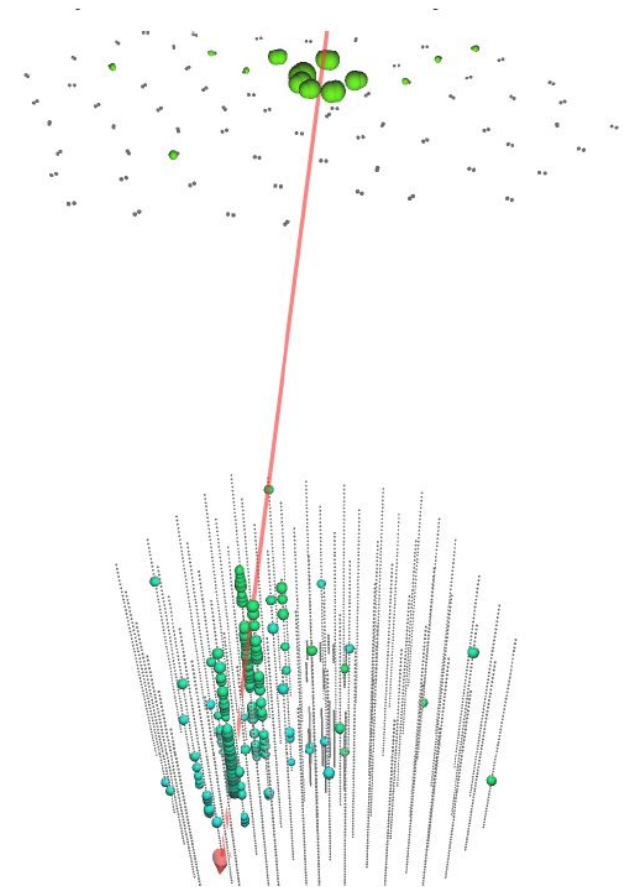
- Production of air showers:
  - gamma-ray primary
  - hadronic primary
- Gamma-ray induced air showers:
  - fewer muons
  - less shower fluctuations
  - narrower lateral spread
- Hadronic air showers:
  - richer in muon content
  - more shower fluctuations
  - wider lateral spread



Images source: Multi-year Campaign of the Gamma-Ray Binary LS I +61° 303 and Search for VHE Emission from Gamma-Ray Binary Candidates with the MAGIC Telescopes  
 doi: 10.13140/RG.2.1.4140.4969

# IceCube Neutrino Observatory

- Ice-Top:
  - 1 km<sup>2</sup>
  - 162 Cherenkov tanks
  - measurement of the electromagnetic component of the shower
- In-ice:
  - 1 km<sup>3</sup>
  - 86 strings with in-ice with 60 DOMs each
  - measurement of the high energy muons

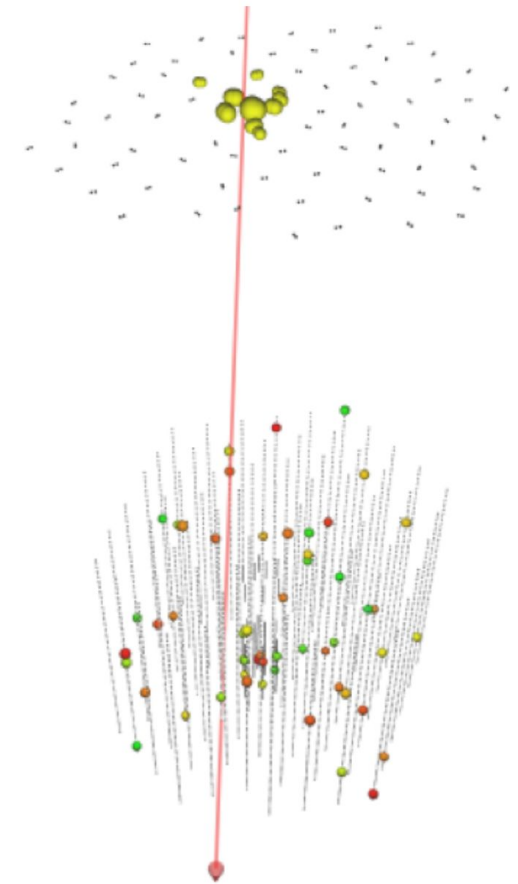


A cosmic-ray air shower event with a shower axis that passes through both components of the detector. (Z. Griffith)

Source: SEARCH FOR PEV GAMMA RAYS WITH THE ICECUBE OBSERVATORY, Zachary Dean Griffith

# IceCube Neutrino Observatory

- Ice-Top:
  - 1 km<sup>2</sup>
  - 162 Cherenkov tanks
  - measurement of the electromagnetic component of the shower
- In-ice:
  - 1 km<sup>3</sup>
  - 86 strings with in-ice with 60 DOMs each
  - measurement of the high energy muons



An event from the 2012 point source sample dataset with one of the ten highest Random Forest scores. (Z. Griffith)

Source: SEARCH FOR PEV GAMMA RAYS WITH THE ICECUBE OBSERVATORY, Zachary Dean Griffith

# Find Gamma ray sources

1. Data (burning sample) and Sim (Sibyll 2.1) .i3 -> .hdf5
2. Cleaning & selection
3. Train Random Forest
4. Data selection with the Random Forest -> Gamma selection
5. Background creation
6. All sky search
7. Test statistic
8. P-value
9. Final plot

Completed for year 2012, 2013, 2014 and 2015  
burning sample year 2011 is now available

Simulations:

[https://wiki.icecube.wisc.edu/index.php/Cosmic-ray\\_IC86\\_Datasets](https://wiki.icecube.wisc.edu/index.php/Cosmic-ray_IC86_Datasets)

# Data and Sim Cleaning & selection

- Quality cuts:
  - Number of triggered Station  $\geq 5$
  - Fraction Containment  $< 1.0$  (contained in the in the surface array)
  - Laputop zenith  $< \arccos(0.8)$
  - $\log_{10}(S125) > -0.25$
  - Laputop Beta  $> 1.4$  & Laputop Beta  $< 9.5$
  - $\log_{10}(\text{Energy} / \text{GeV}) < 8.0$
  
- Features for Random Forest:
  1. charges: total in-ice charge
  2. Laputop in-ice FractionContainment
  3. Laputop  $\log_{10}(s125)$
  4.  $\sin(\text{dec})$
  5. LLH\_Ratio: log-likelihood parameter (Hershal PhD thesis)

# Log likelihood ratio

The log-likelihood ratio is constructed for each event via 3 PDFs

e.g. The log-likelihood ratio for charge Q and lateral distance R:

$$\Lambda_{QR} = \log_{10} \left( \frac{L_{QR}(\text{event}|H_{\gamma})}{L_{QR}(\text{event}|H_{CR})} \right), \quad L_{QR}(\text{event}|H) = \prod_{i=1}^{162} P(Q_i, R_i|H),$$

with  $P(Q_i, R_i | H)$  being the probability of observing a tank with measured charge  $Q_i$  and at lateral distance  $R_i$ , for the hypothesis  $H$ .

Total log-likelihood ratio is defined by the sum of all three log-likelihood ratios

$$\Lambda = \Lambda_{QR} + \Lambda_{Q\Delta T} + \Lambda_{\Delta TR}$$



# Random Forest

The Random Forest: open-source Python package Scikit-Learn

A Random Forest classifier consists of a combination of many simple decision trees

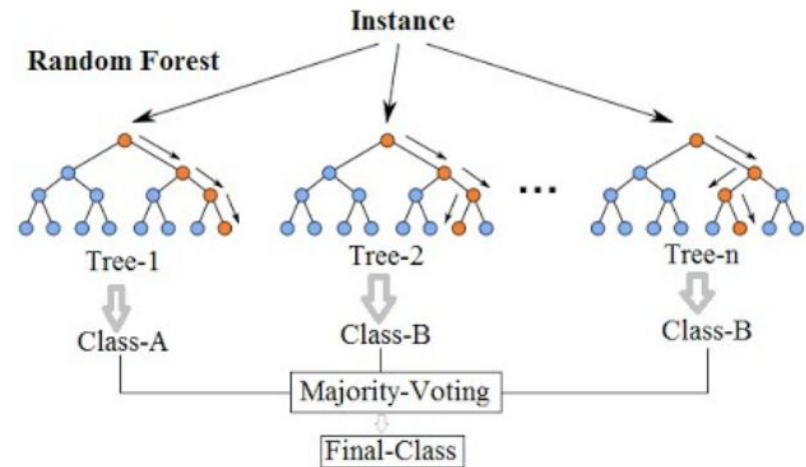
Each tree acquires its n events from the training sample, such that each tree trains on a different set of events every time

The splitting condition: minimization of the Gini impurity in the child nodes

$$I_G = 1 - \frac{w_S^2 + w_B^2}{(w_S + w_B)^2}$$

$w_S$  and  $w_B$  are the total weights in a node for the signal and background classes

“probability of misclassification”:  
 all of the weight in the node is in one class  $I_G = 0$   
 an even split in the node results in  $I_G = 0.5$ .



Source: <https://scikit-learn.org/stable/>

# Random Forest

The Random Forest: open-source Python package Scikit-Learn

A Random Forest regressor consists of a combination of many simple decision trees

The splitting condition: minimization of the Mean Square Error (MSE)

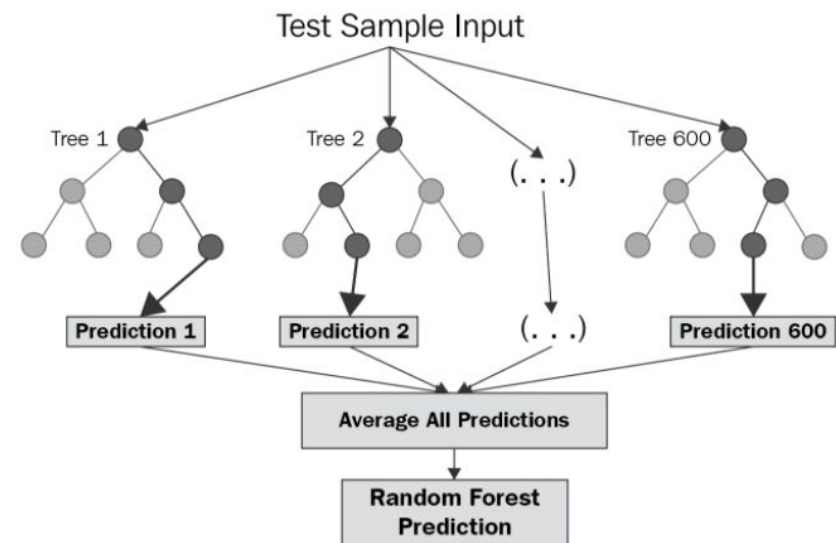
$$\text{MSE} = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

MSE = mean squared error

$n$  = number of data points

$Y_i$  = observed values

$\hat{Y}_i$  = predicted values



Source: <https://scikit-learn.org/stable/>

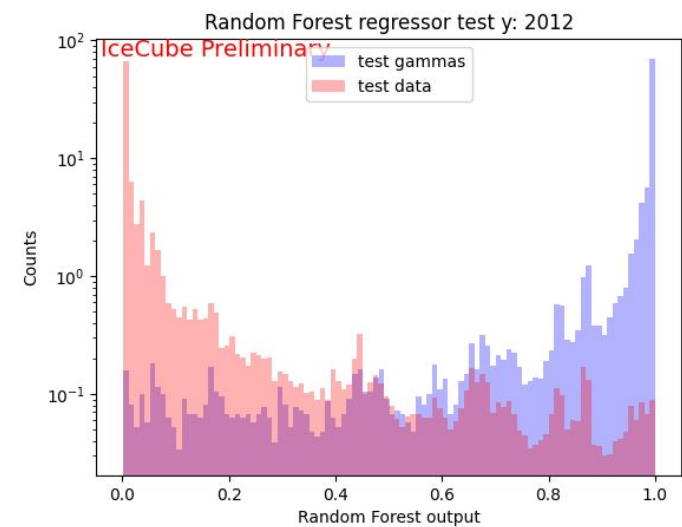
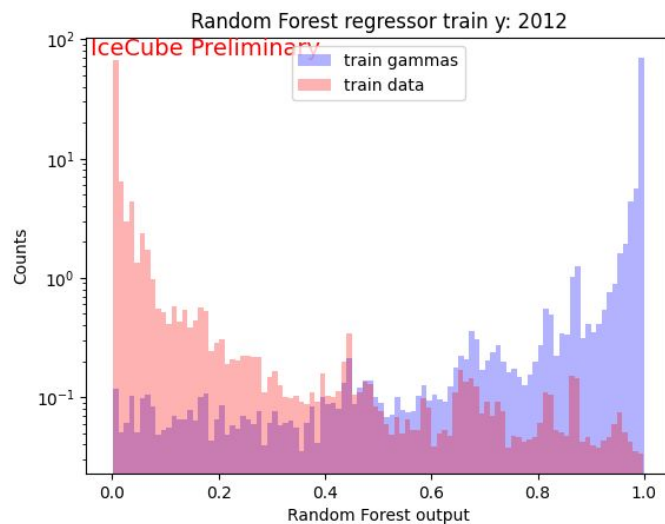
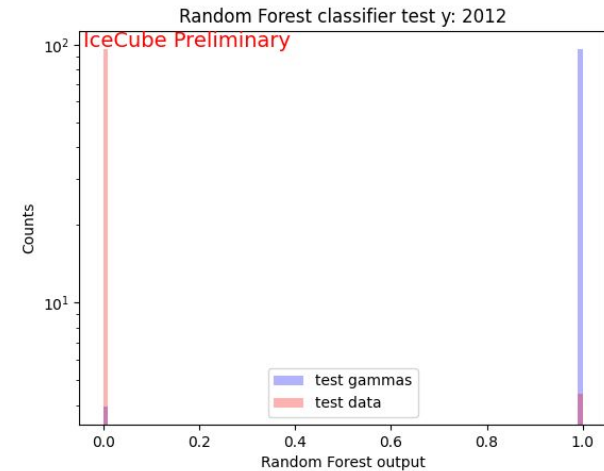
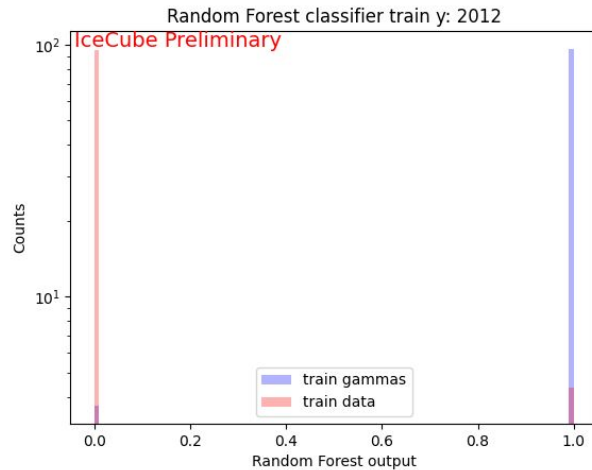
# Training & Testing

- Training sample:
  - about 80.000 Gamma MC
  - about 80.000 Data as background
- Test sample:
  - about 20.000 Gamma MC
  - the complete burn sample about 4 Millions event
- Note: a different forest was trained for each year (snow accumulation)

```
RandomForestClassifier(n_estimators=100,                                cut value=0.7  
                       random_state=0,  
                       n_jobs=5, verbose=0,  
                       max_depth=6)
```

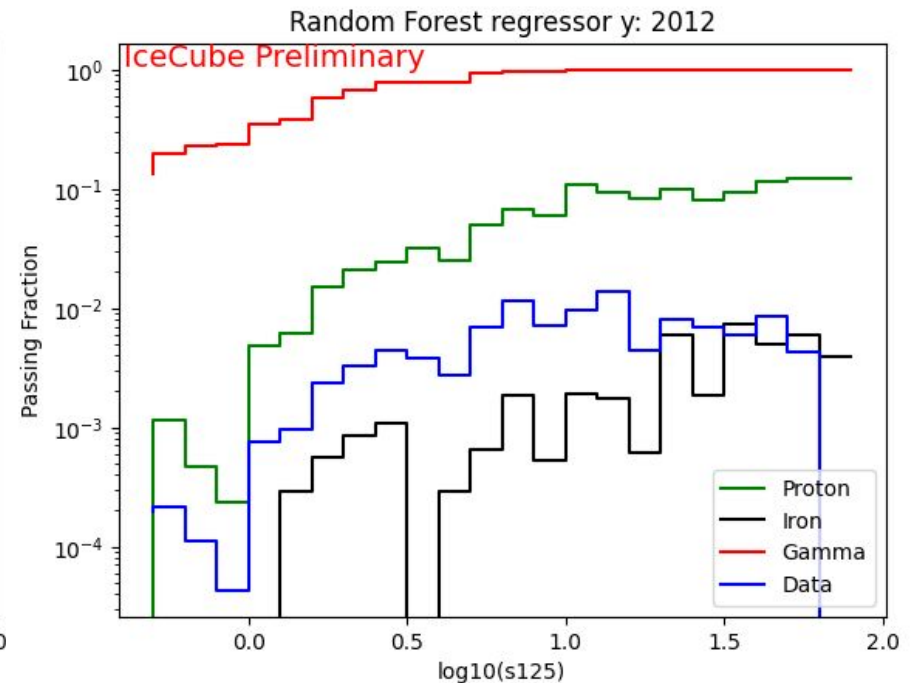
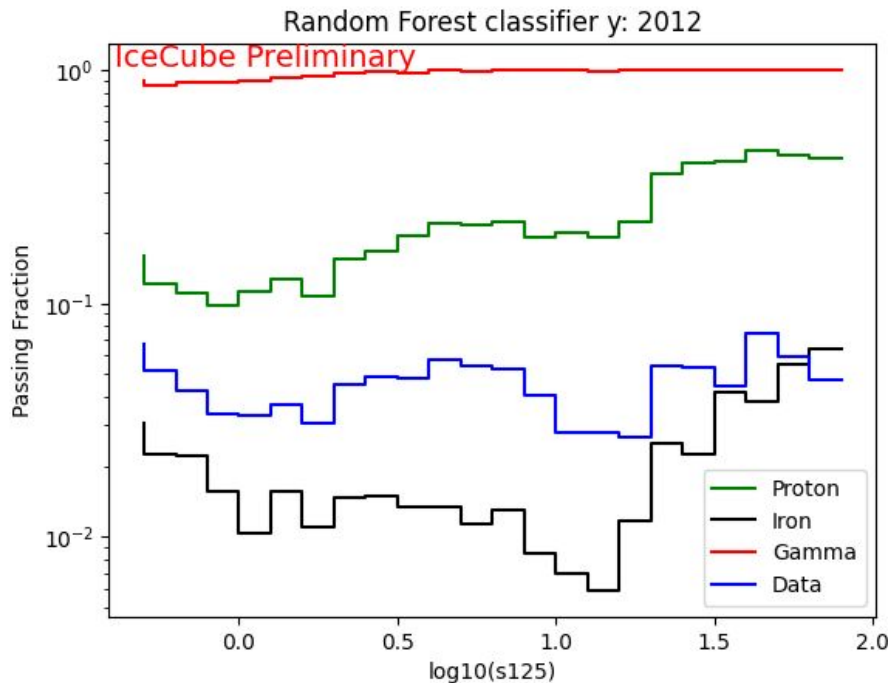
```
RandomForestRegressor(n_estimators=100,                               cut value=0.99  
                      random_state=0,  
                      n_jobs=5, verbose=0,  
                      max_depth=8)
```

# Training & Testing year 2012



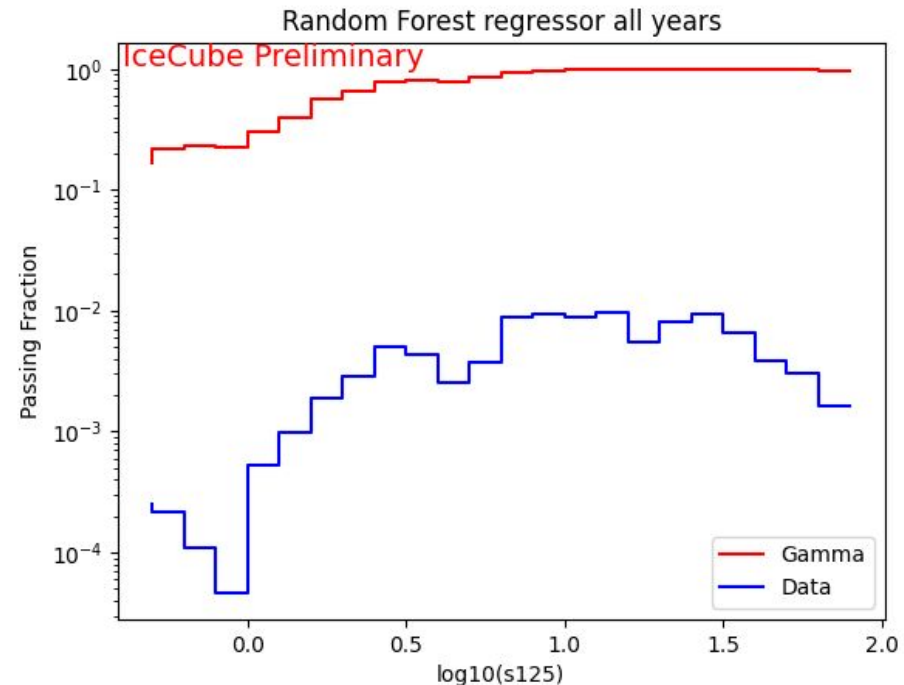
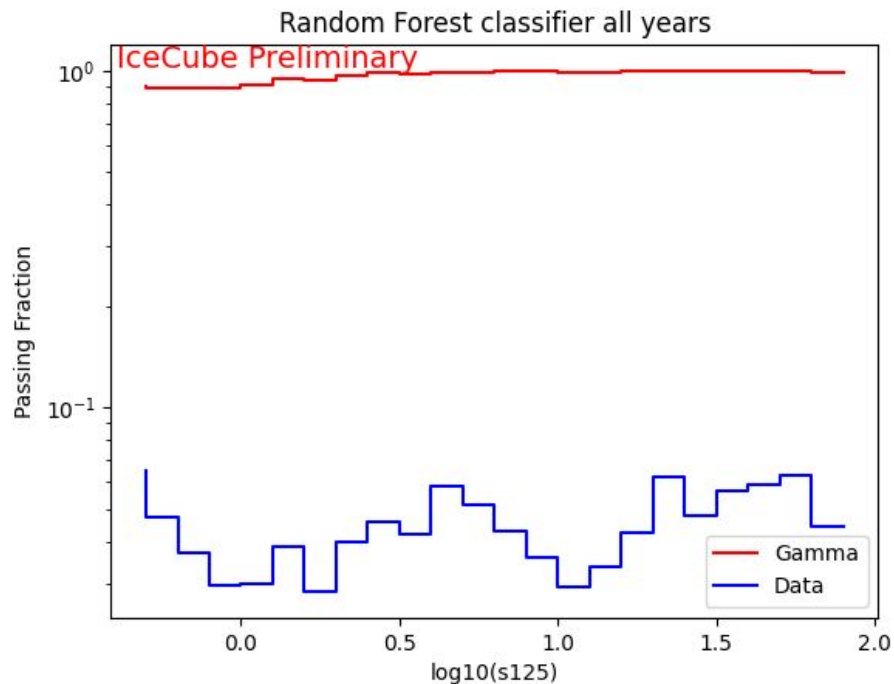
# Passing fraction year 2012

- event that passed the prediction cut / Total event
- plotted as function of energy



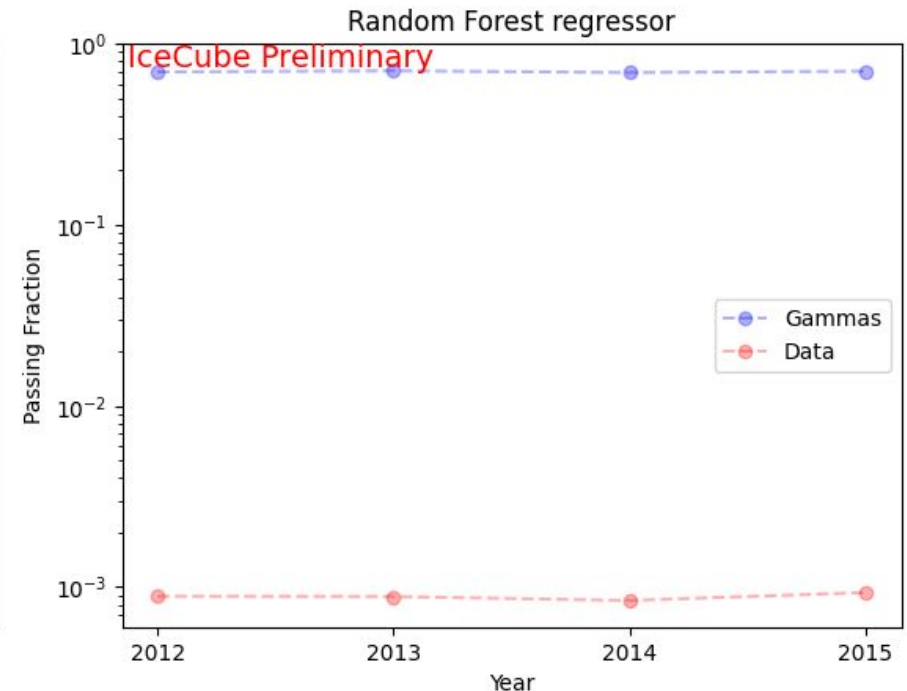
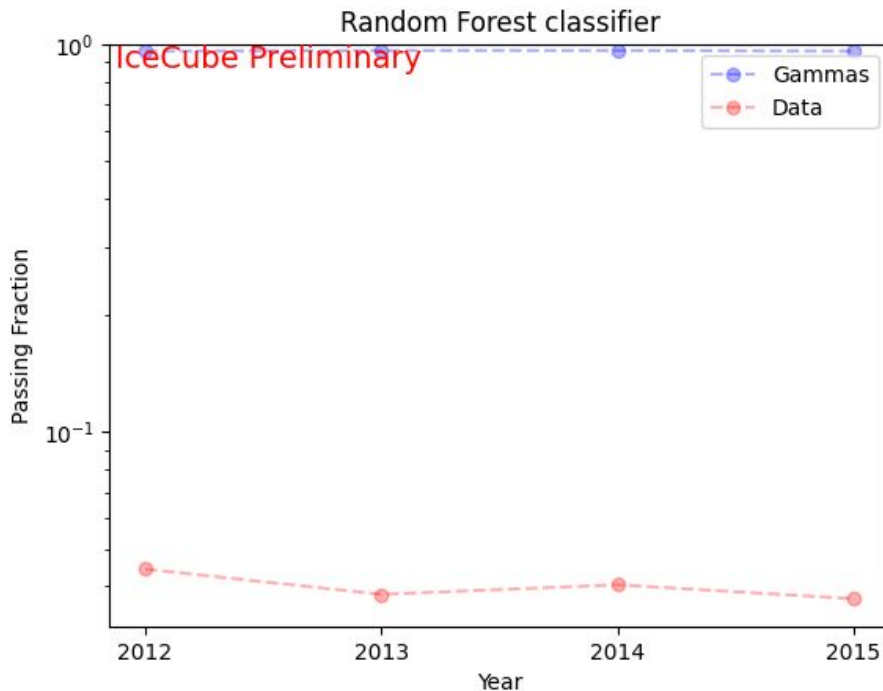
# Passing fraction all years

- event that passed the prediction cut / Total event
- plotted as function of energy



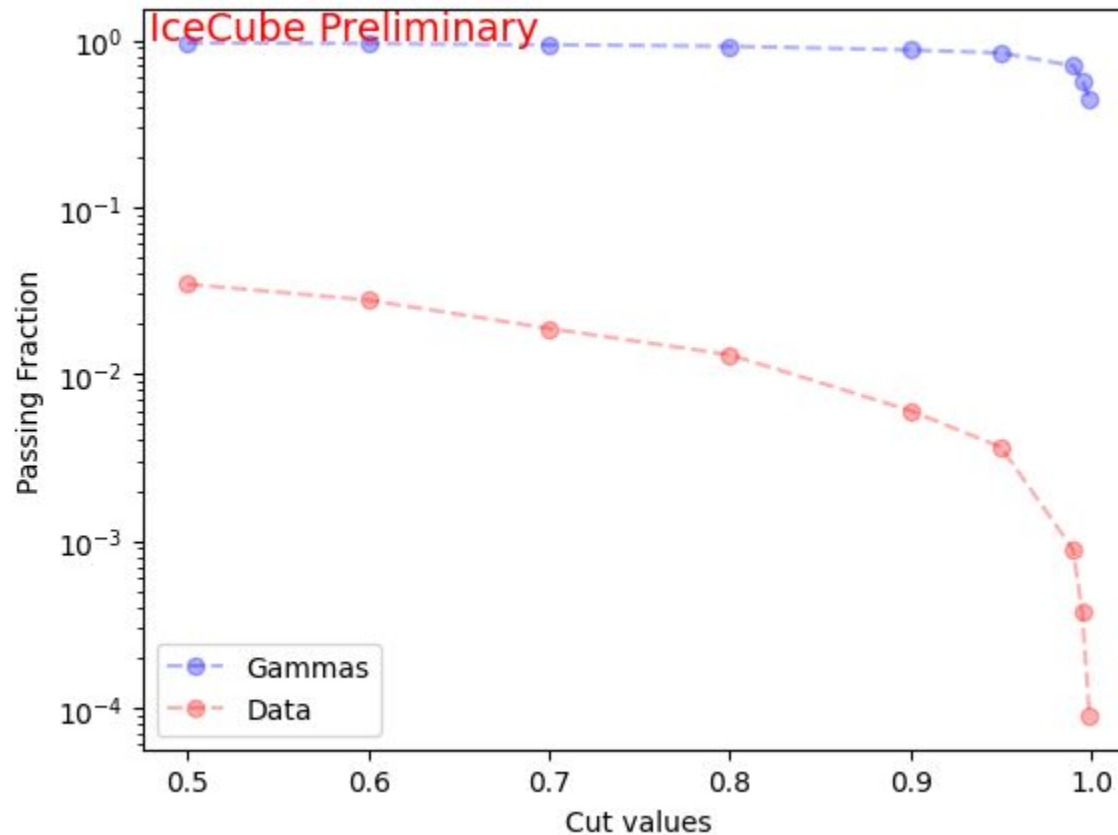
# Passing fraction all years

- event that passed the prediction cut / Total event



# Passing fraction regressor all years

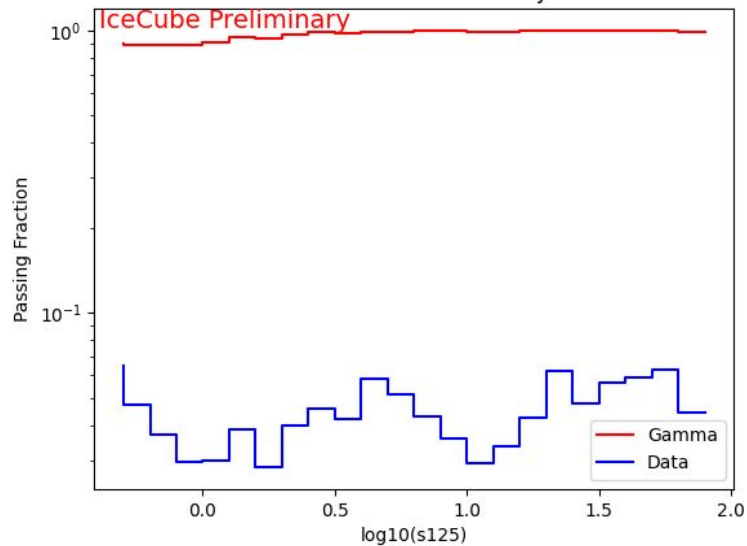
- Passing fraction for different cut values:  
[0.5,0.6,0.7,0.8,0.9,0.95,0.99,0.995,0.999]



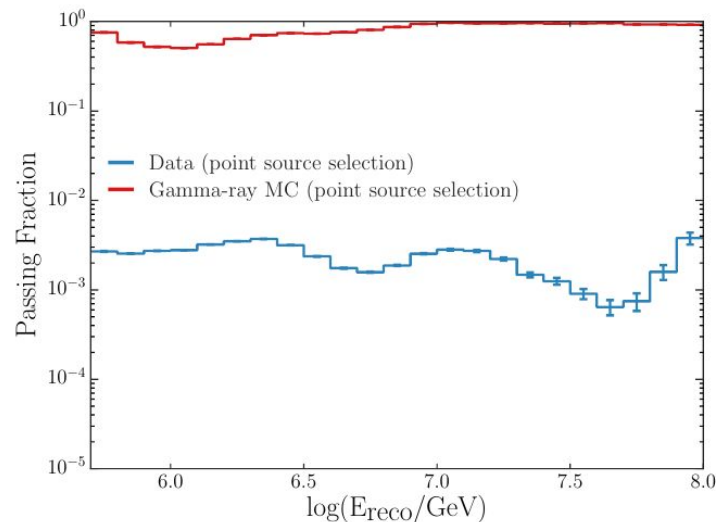
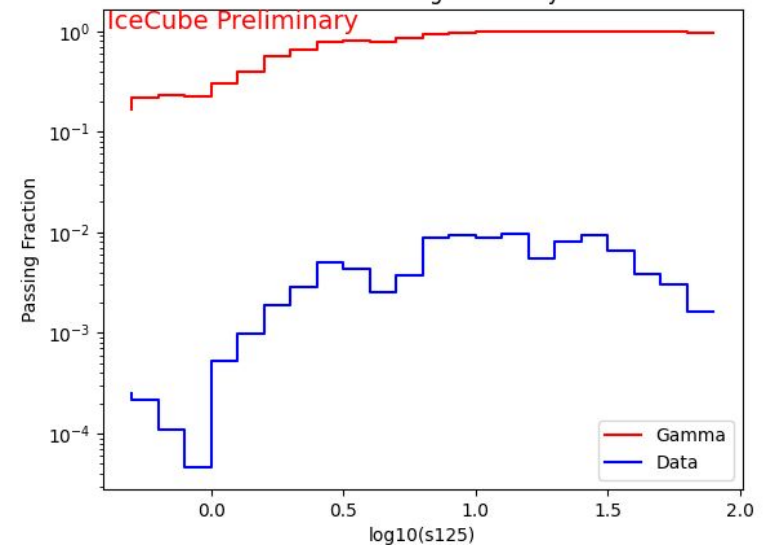


# Comparison with Zach's results

Random Forest classifier all years



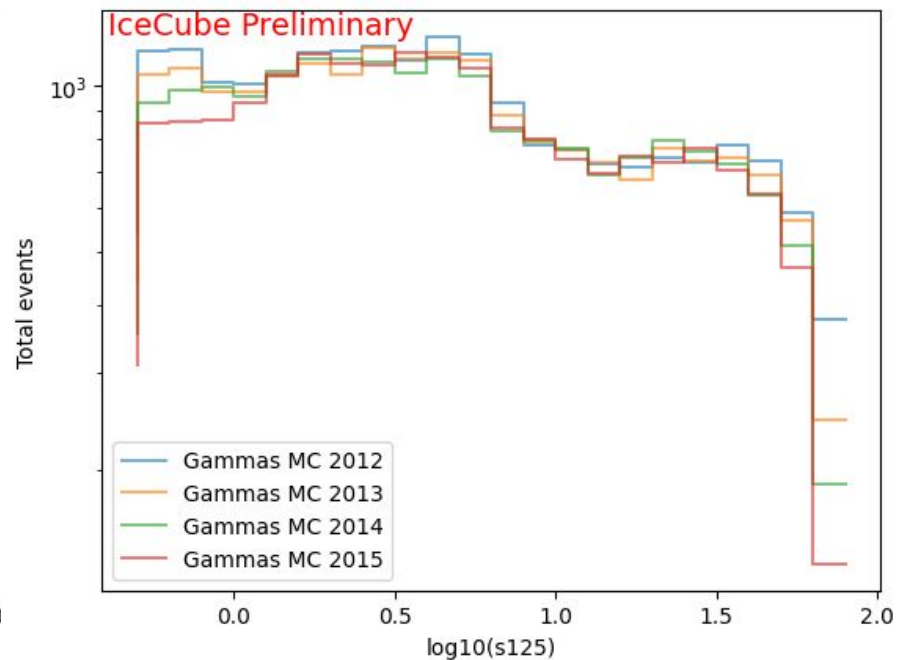
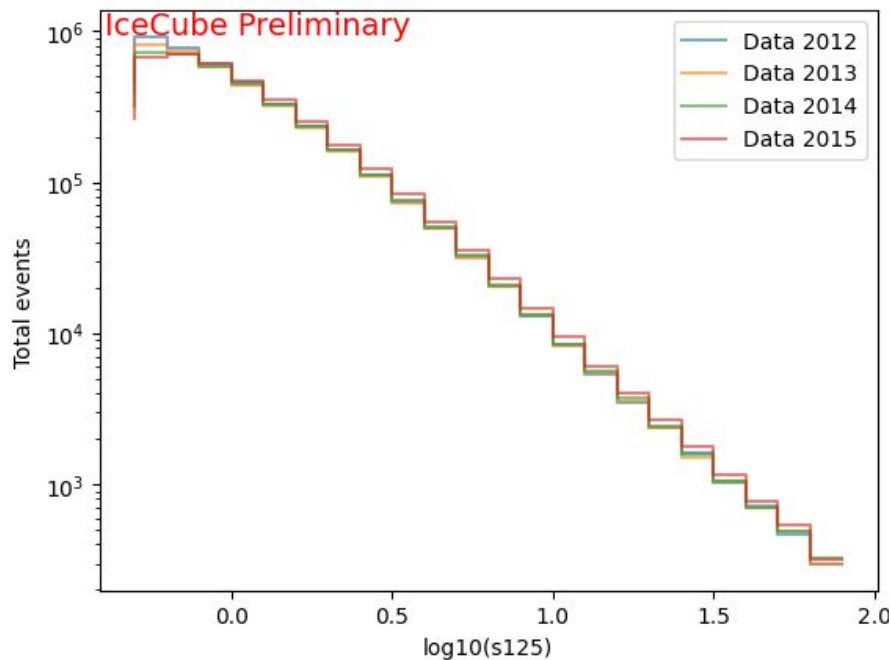
Random Forest regressor all years



Source: SEARCH FOR PEV  
GAMMA RAYS WITH THE  
ICECUBE OBSERVATORY,  
Zachary Dean Griffith

# Data and Sim Cleaning & selection

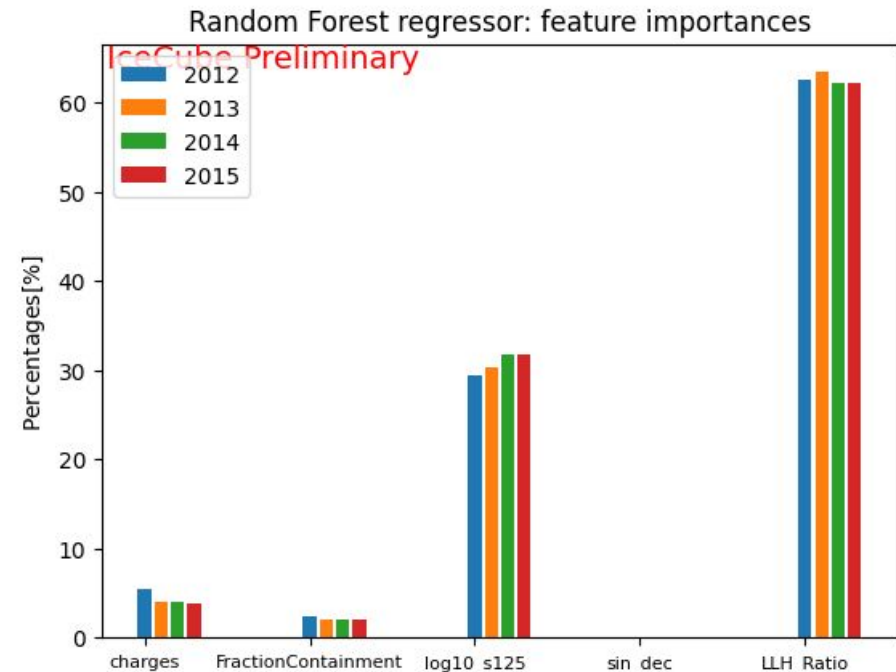
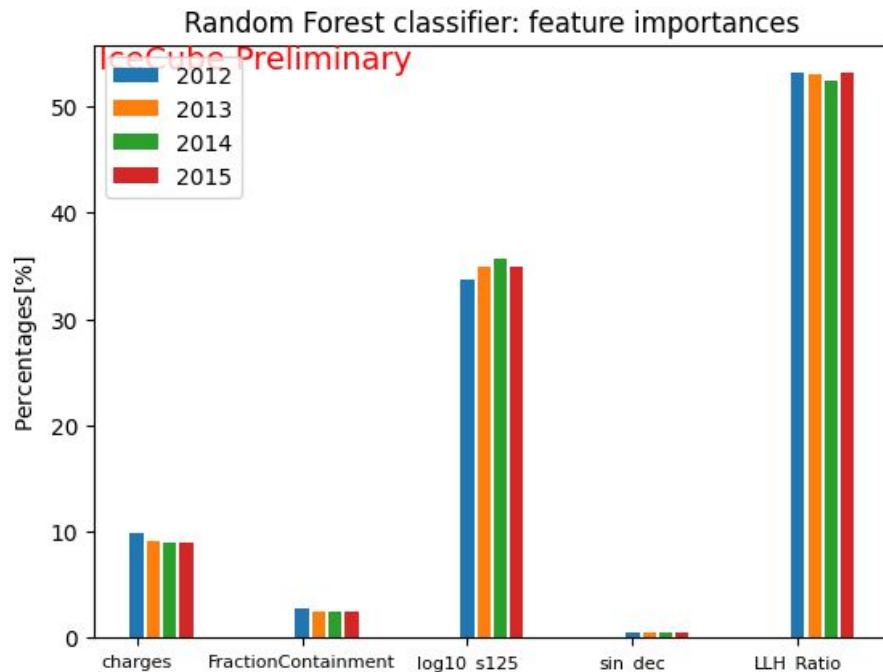
Sim and Data do not have the same steepness



# Feature importance all years

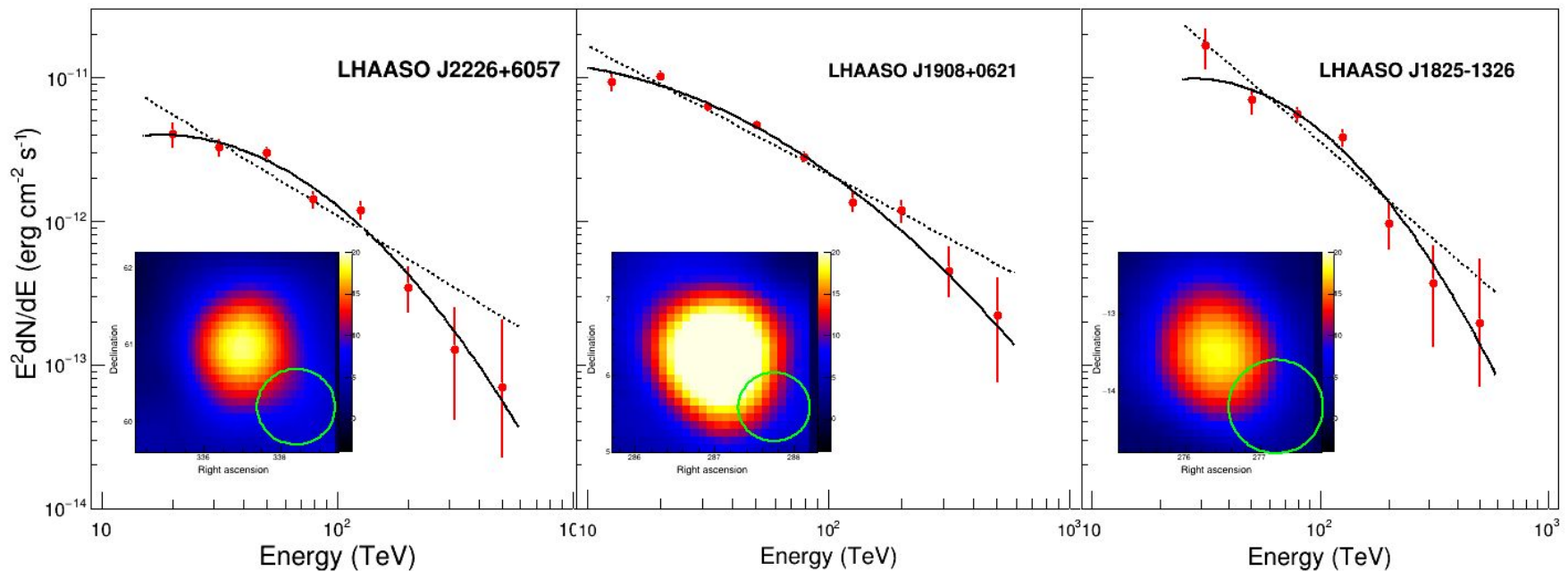
Each of the input feature has a different importance.

For each year, here they are plotted in percentages and sum up to 1



# LHAASO sources

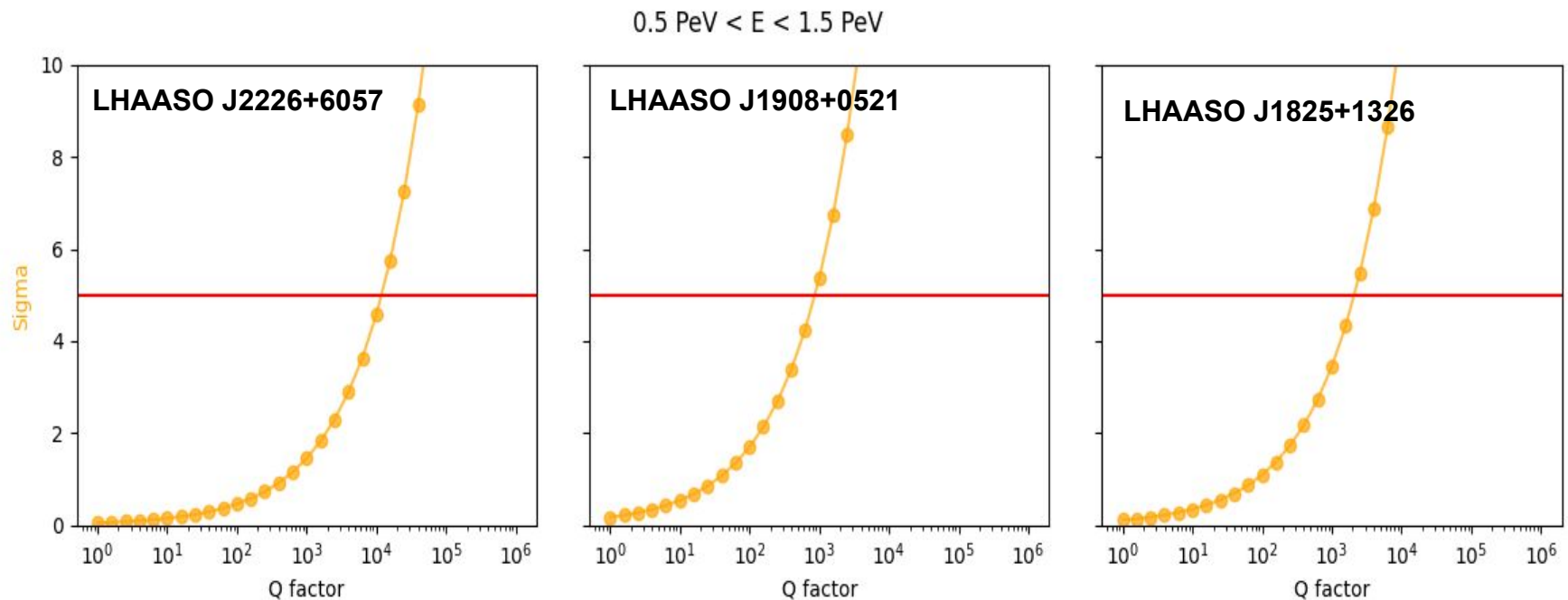
- Ultra High-energy photons up to 1.4 petaelectronvolts from 12  $\gamma$ -ray Galactic sources
- Is IceCube able to detect a LHAASO like UHE photon source?



Source: <https://www.nature.com/articles/s41586-021-03498-z>

# Sigma and Q-factor

- Q-factor indicates how good the CR background can be rejected
- A Q-factor between  $10^3 \div 10^4$  is required for signal bg separation for a 5 sigma observation
- # sigma = signal / bg\*\*0.5



# Summary & Outlook

- The classifier and the regressor are two different approaches
- Better gamma hadron separation for lower energies probably due to lower statistics for data at higher energies
- The declination is almost irrelevant for the final selection
- Regressor and Classifier depend on the cut value selected

Coming next:

- Correct the energy bias between simulation and data
- Search for gamma sources in the sky
- Use Deep Learning neural networks for the classification

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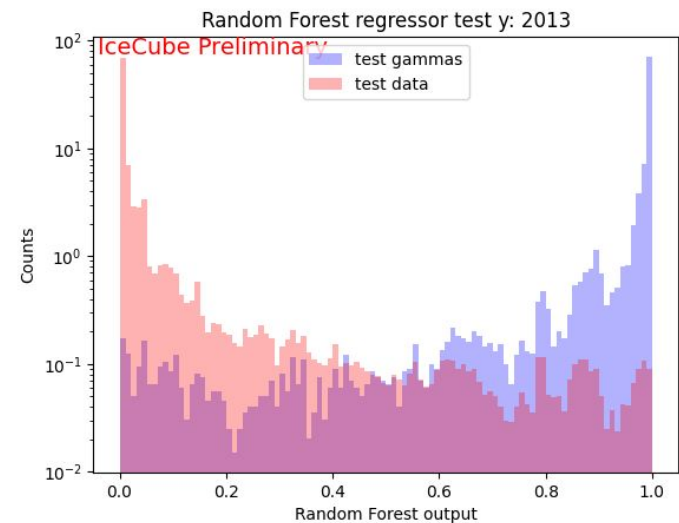
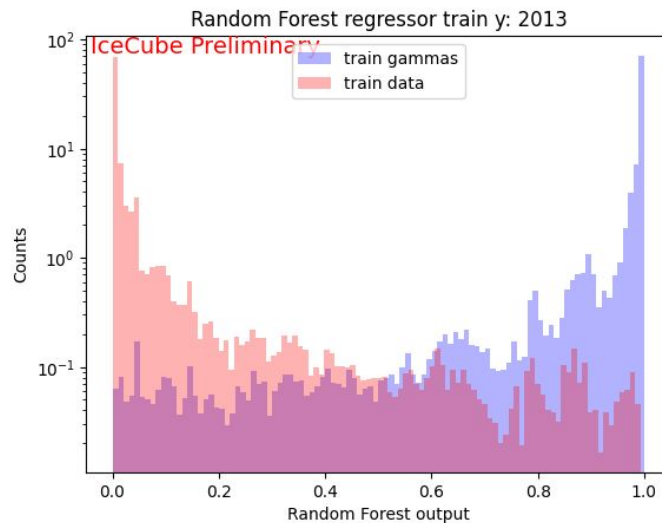
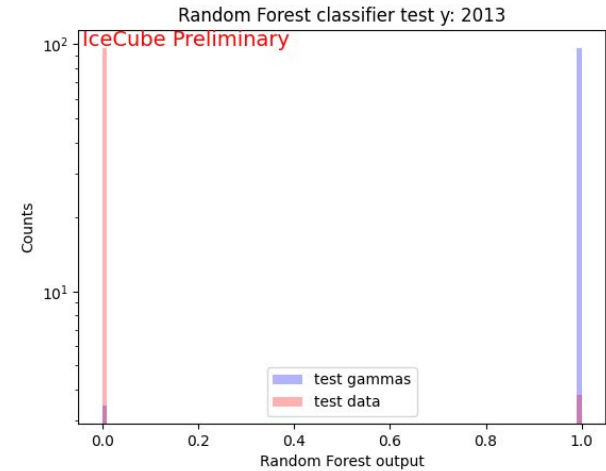
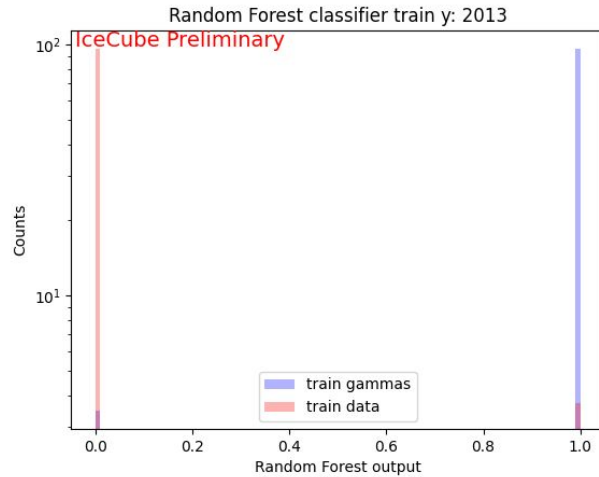


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SOUTH POLE NEUTRINO OBSERVATORY

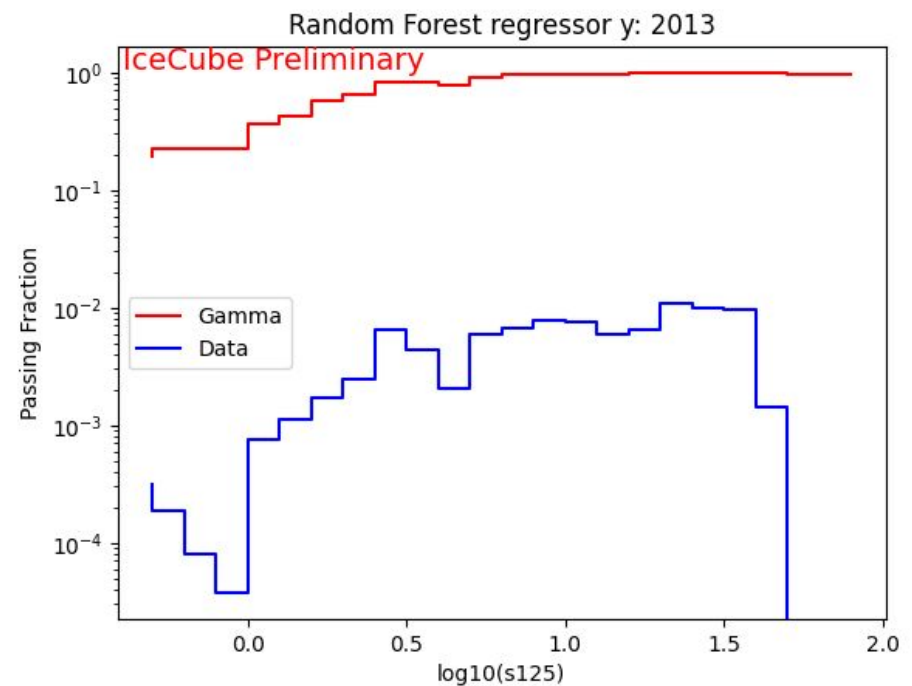
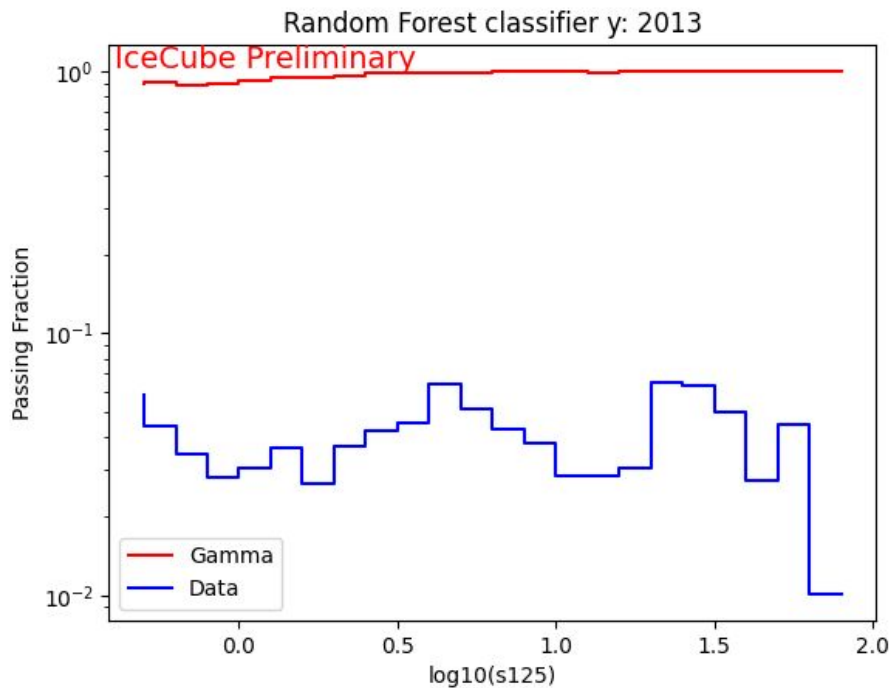
# Training & Testing year 2013



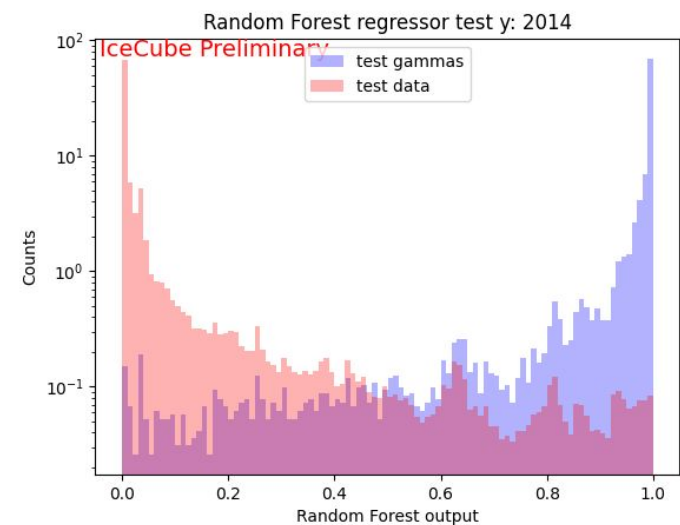
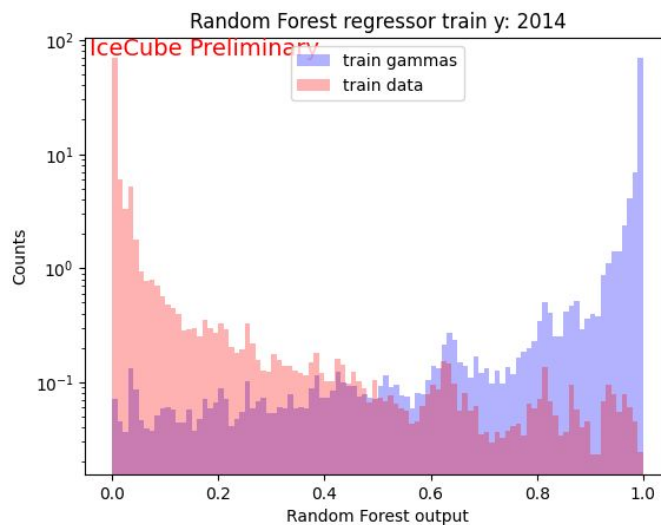
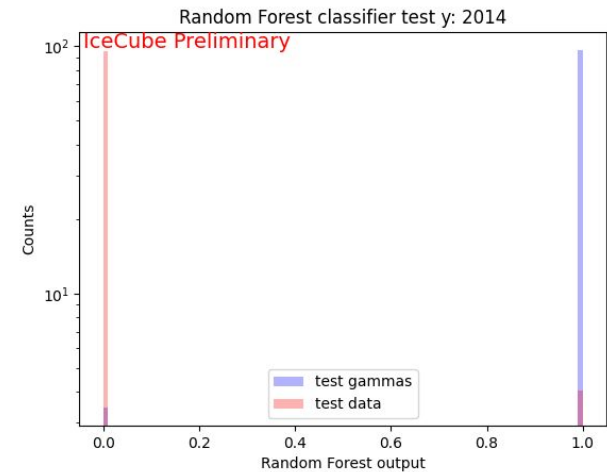
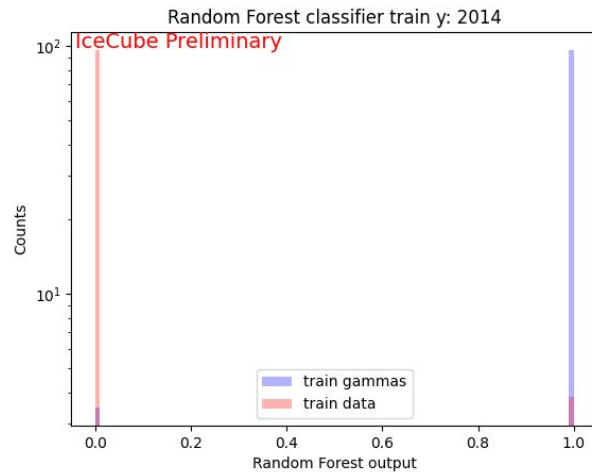


# Passing fraction year 2013

- event that passed the prediction cut / Total event
- plotted as function of energy

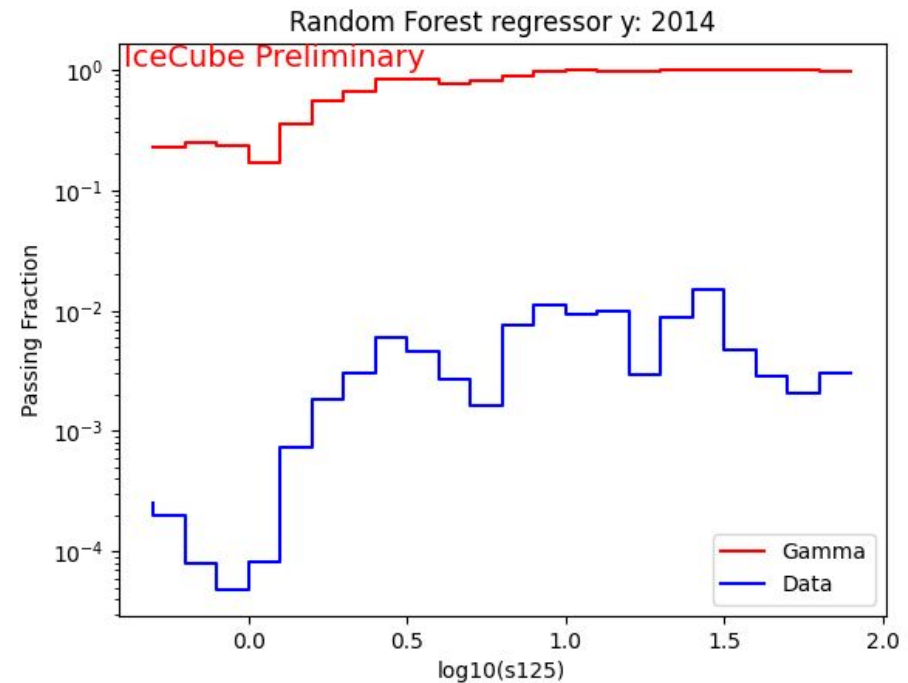
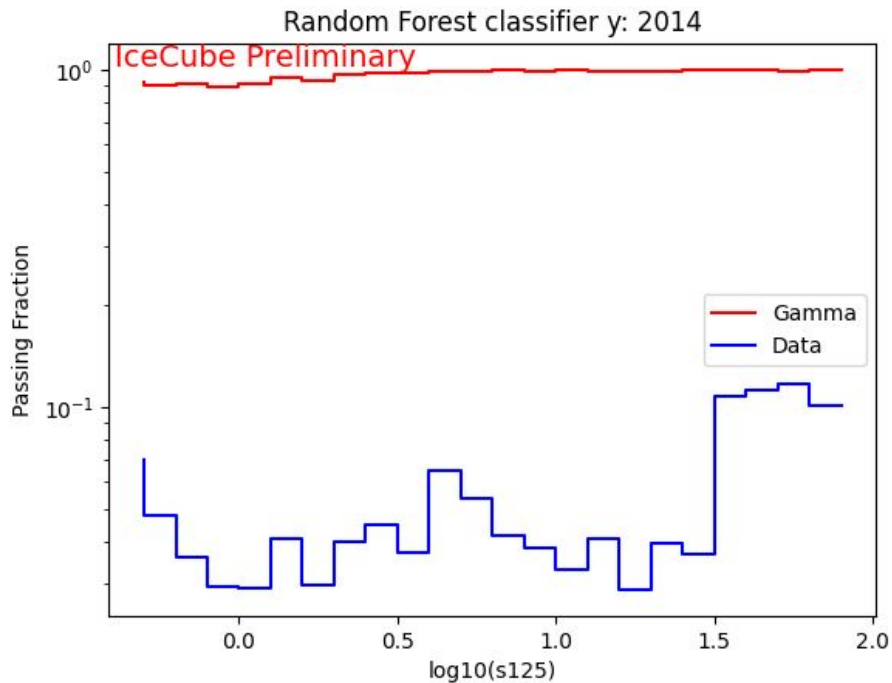


# Training & Testing year 2014

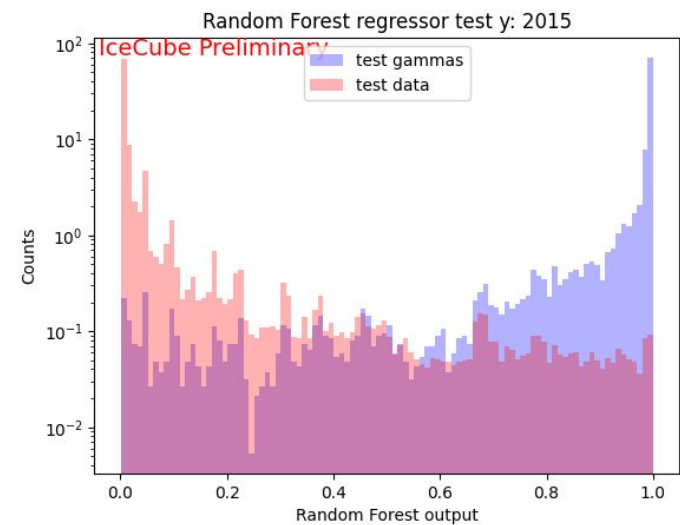
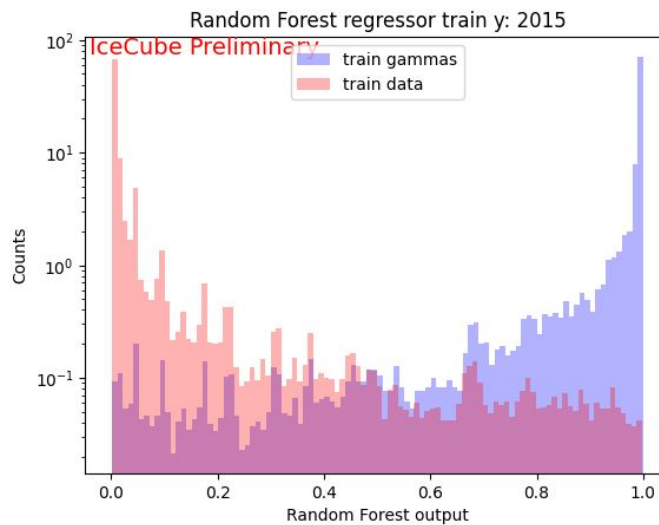
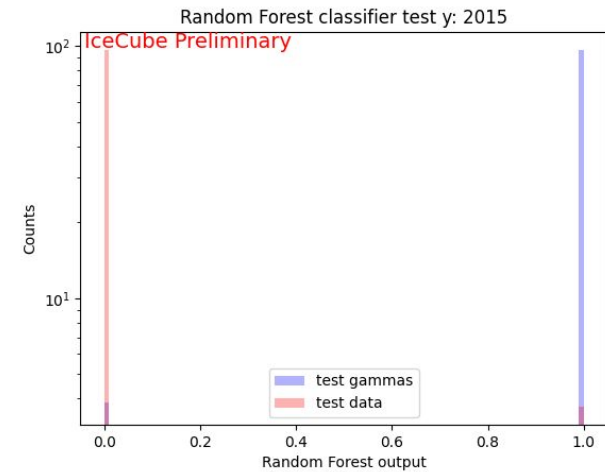
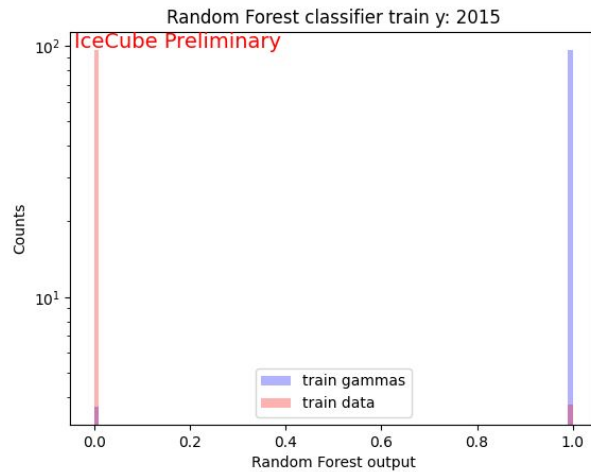


# Passing fraction year 2014

- event that passed the prediction cut / Total event
- plotted as function of energy

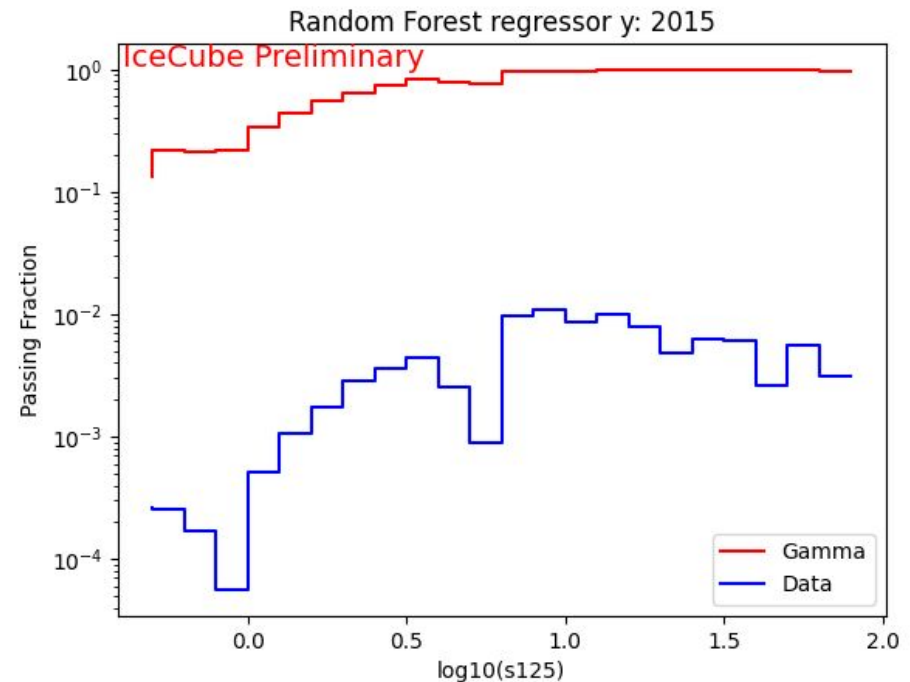
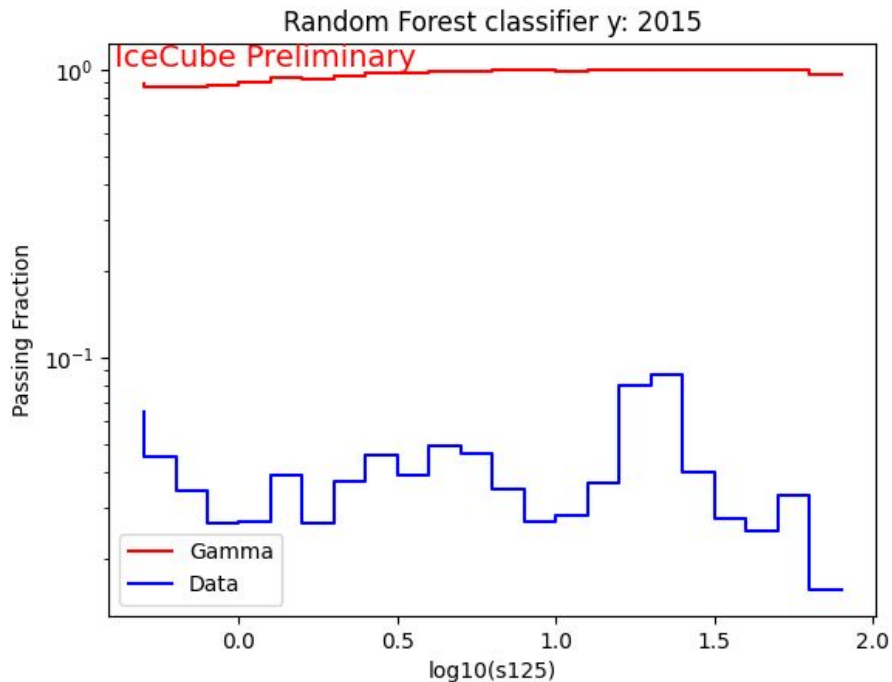


# Training & Testing year 2015



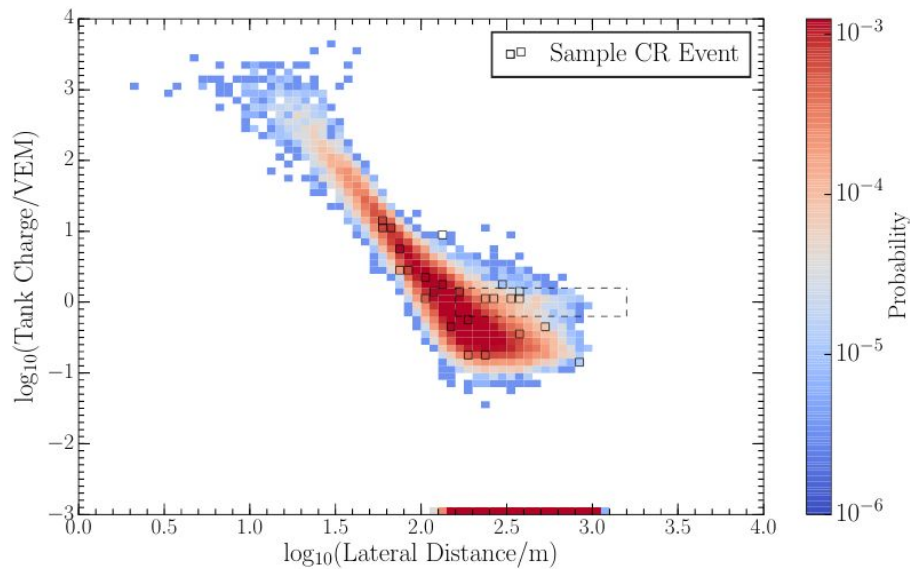
# Passing fraction year 2015

- event that passed the prediction cut / Total event
- plotted as function of energy

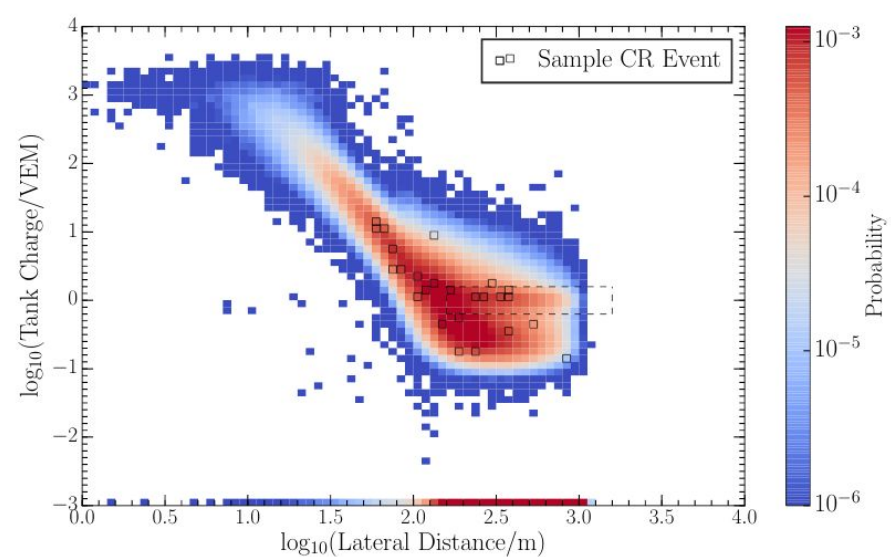


# Log likelihood ratio

- example of a photon and CR background probability map



(a) Lateral PDF for gamma rays.



(b) Lateral PDF for cosmic rays.

Source: SEARCH FOR PEV GAMMA RAYS WITH THE ICECUBE OBSERVATORY, Zachary Dean Griffith