

# GEP qq Tagger

Santosh Parajuli

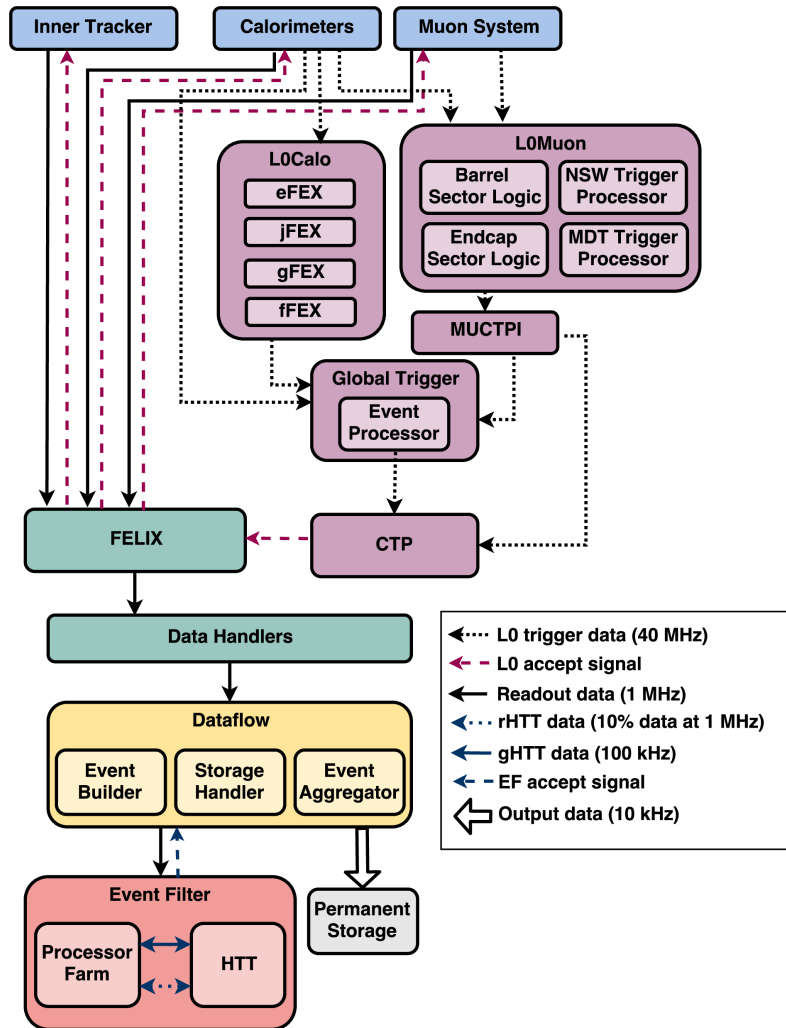
**Southern Methodist University**

**1 March 2023**

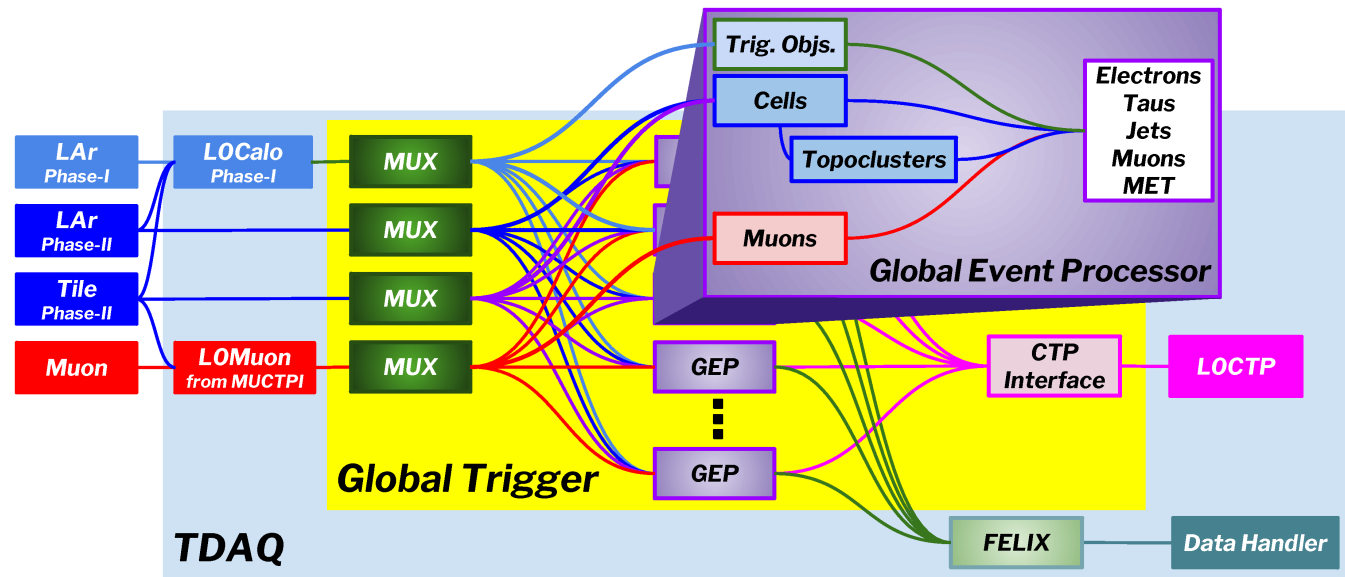


**SMU**<sup>®</sup>

# Global Trigger and Global Event Processor(GEP) in ATLAS



TDAQ System after the Phase-II upgrade (baseline design)



Schematic view of the Global Trigger System

# Quark/Gluon Tagger Algorithm for GEP

- Binary Classifier Implementation, CNN Model.
- GEP Simulation Sample is used.

**1) qcd-multijet sample-> ntuple with the branches,**

```
[b'AntiKt10LCTopoLeadJets_partonTruthLabel',  
b'AntiKt10LCTopoLeadJets_pt',  
b'AntiKt10LCTopoLeadJets_nConstitutents',  
b'AntiKt10LCTopoLeadJets_eta',  
b'AntiKt10LCTopoLeadJets_phi',  
b'AntiKt10LCTopoLeadJets_clus_calE',  
b'AntiKt10LCTopoLeadJets_clus_calEta',  
b'AntiKt10LCTopoLeadJets_clus_calPhi',
```

- Parton truth labels to classify leading  $R=1.0$  jets as quark and gluon.
- Using  $R=1.0$  jets.
- No  $p_T$  reweighting applied. (But the images are normalized)

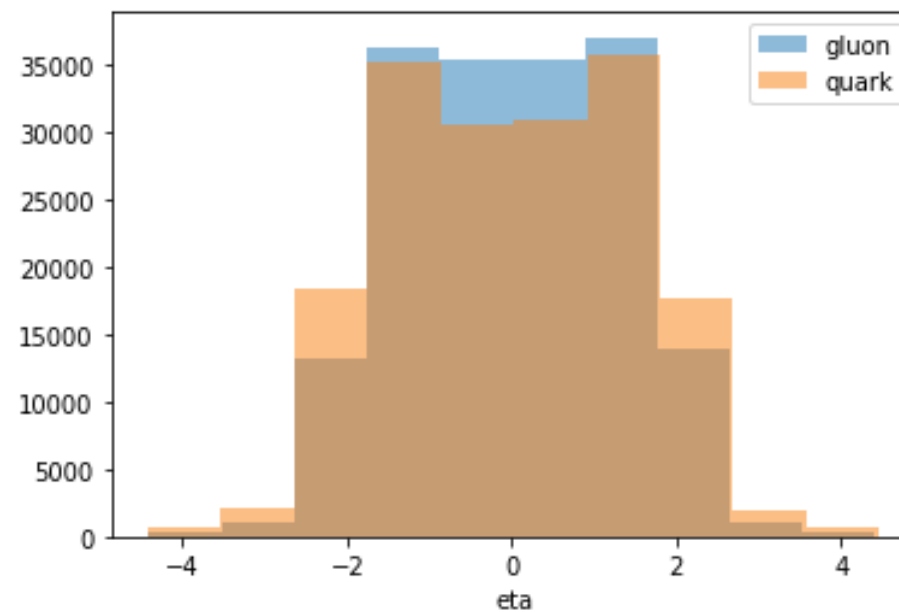
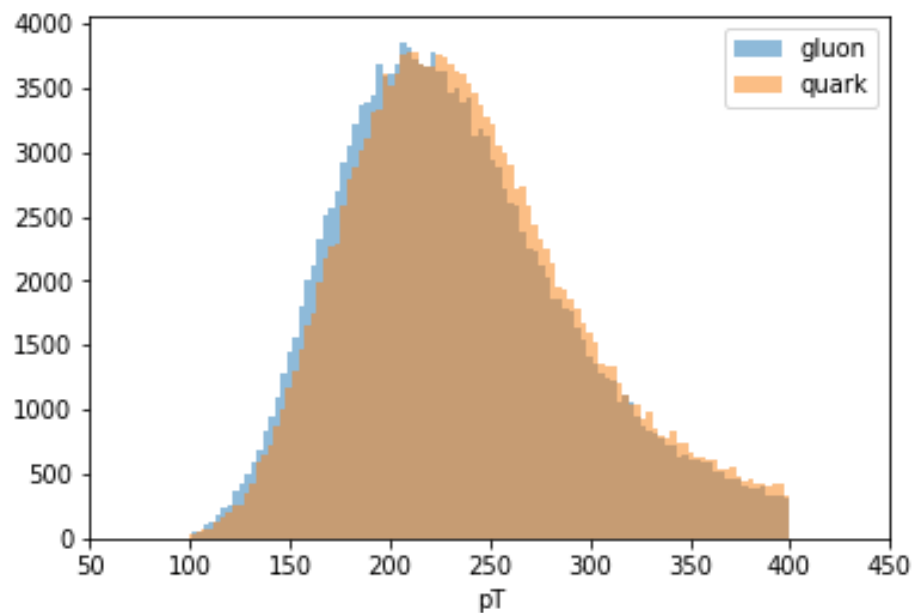
# Motivation for $R=1.0$ jets Quark/Gluon Tagger Algorithm for GEP

- For massive particles that are sufficiently boosted, it is advantageous to reconstruct their hadronic decay products as a single radius (large- $R$ ) jet.
- Hadronically decaying  $W$ ,  $Z$  and  $H$  bosons and top quarks, which is distinct from the radiation pattern of a gluon-initiated jet.
- Searches for heavy resonances.
- Searches for dark matter particles.

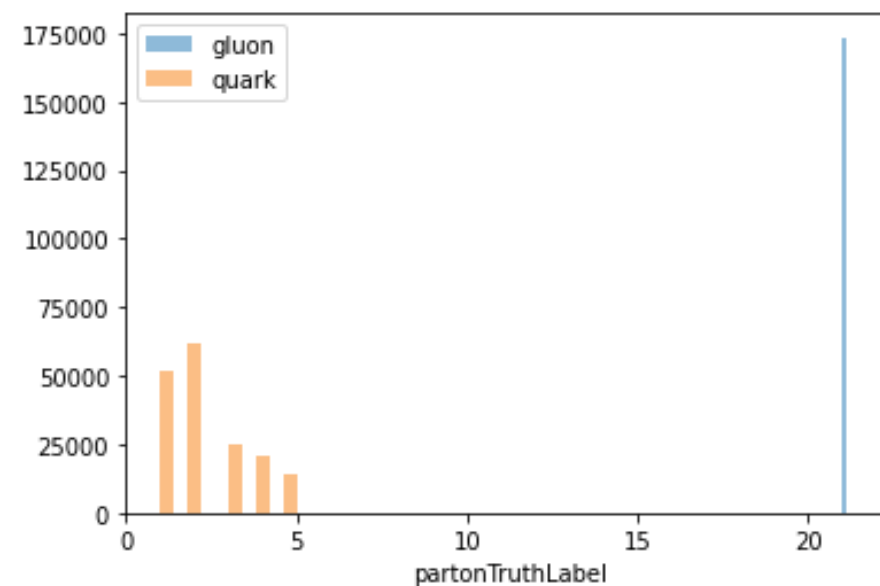
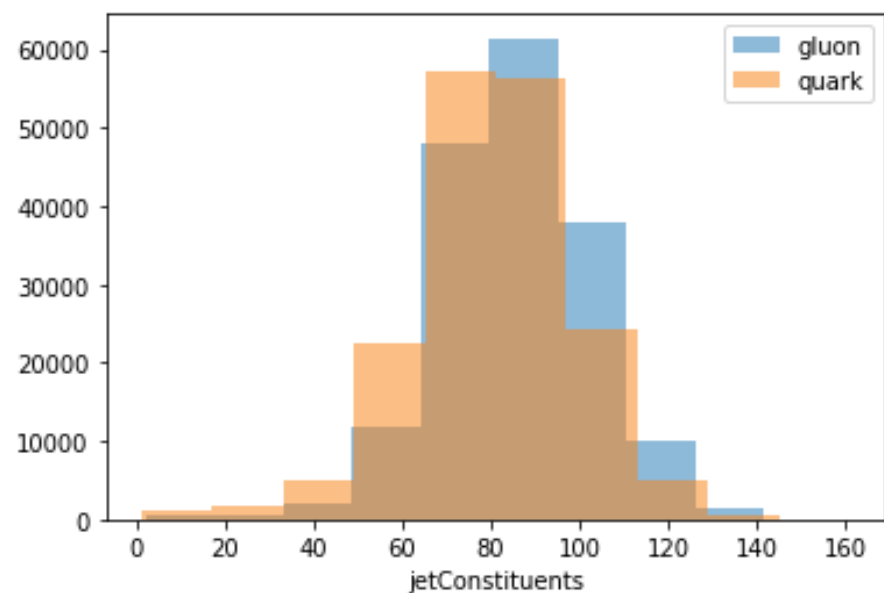
## Some Basic Plots for quark and gluon jets

- **For Quark:** AntiKt10LCTopoLeadJets\_partonTruthLabel  $\geq 1$  and  $\leq 5$  are selected.
- **For Gluon:** AntiKt10LCTopoLeadJets\_partonTruthLabel = 21 is selected.
- $p_T \geq 200$  GeV to  $p_T \leq 250$  GeV events are selected.
- 60,334 signal( quark jets) and 60,334 Background (gluon jets) events are used here. Total-> 120,668 events

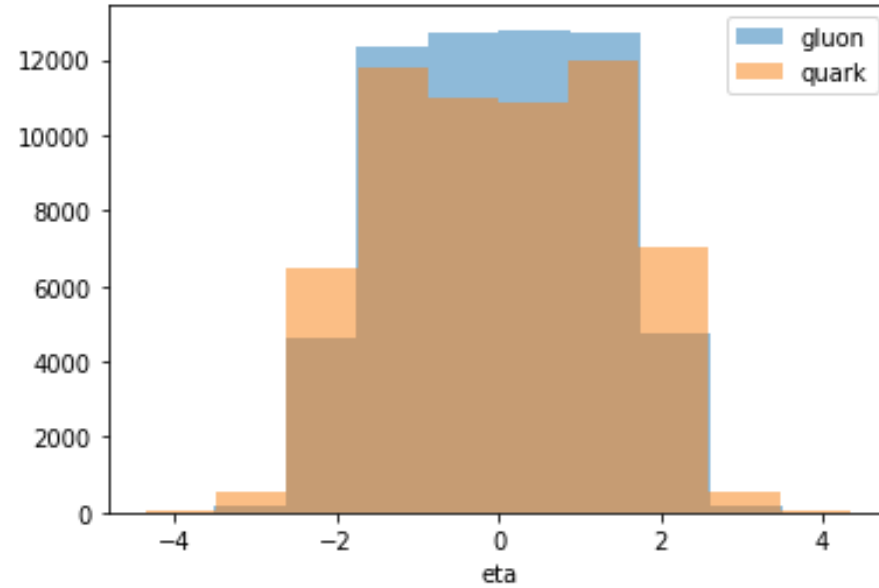
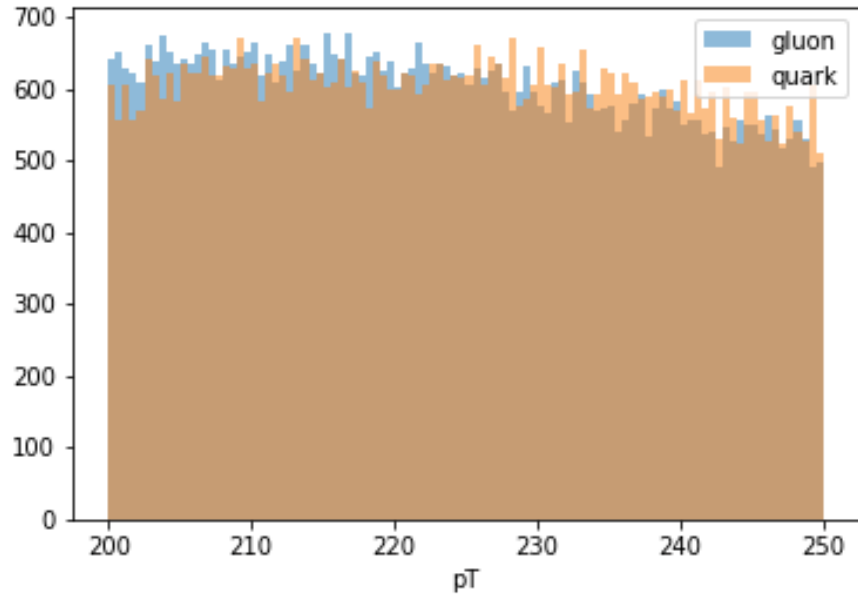
# Some Basic Plots for quark and gluon jets(pT: 100-400 GeV)



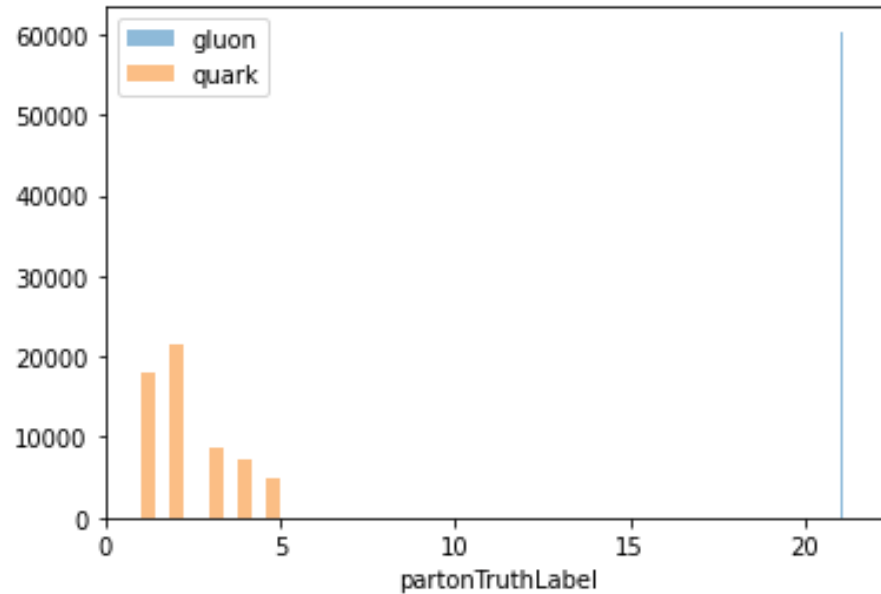
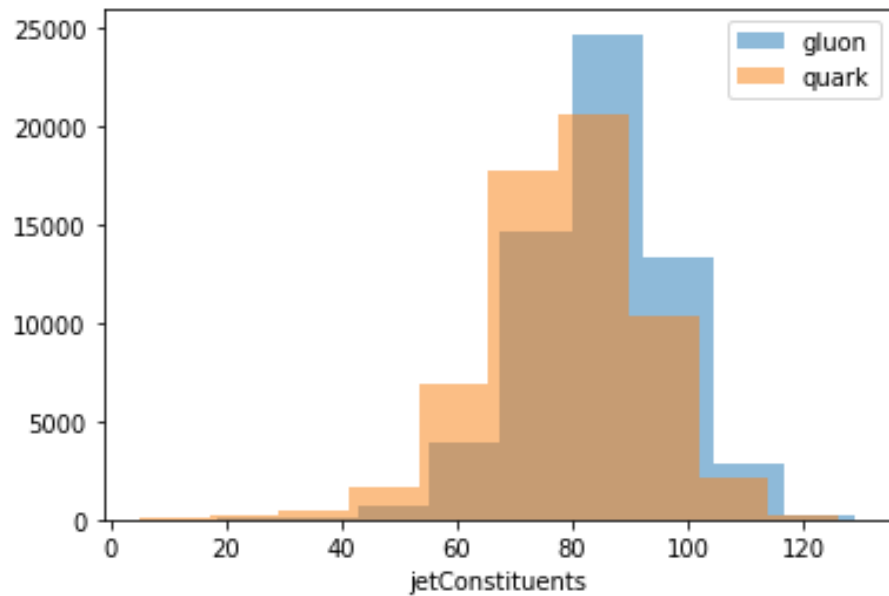
173,461  
events\*2



# Some Basic Plots for quark and gluon jets(pT: 200-250 GeV)



60,334  
events\*2

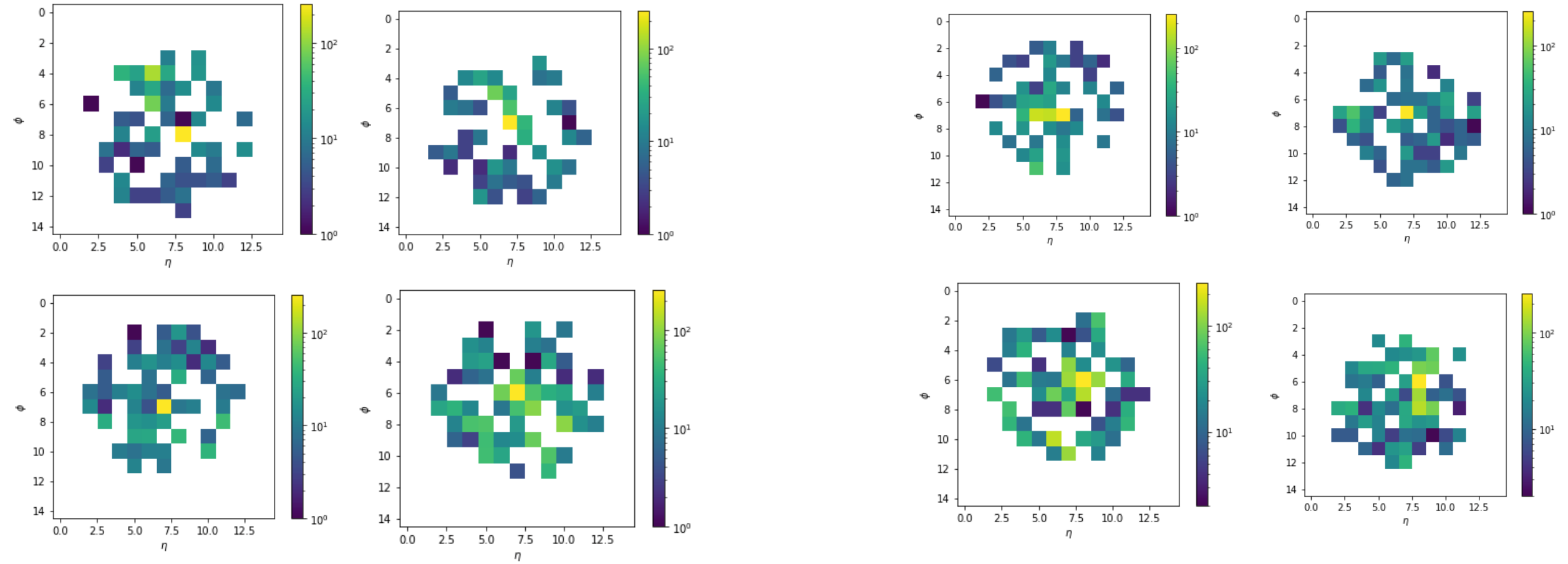


## Making the Jet Image in eta-phi space

- Topological cluster inputs from calorimeter are used to make the jet image.
- The constituents inside a jet are translated in eta and phi so that the jet's center is located at center in eta-phi space.
- 15 \* 15 pixel image in eta-phi space.
- The intensity of each pixel is the total cluster energy within the pixel.
- Pixel value is then normalized by dividing it by the value of the hottest (maximum) pixel in the image. This scaling ensures that the pixel values of the entire image are between 0 and 1. Then, the pixel values are scaled to a range between 0 and 255, this is done by multiplying each pixel value by 255.
- [https://github.com/sparajul/fastmachinelearning/blob/main/preProcessJetImages\\_santosh.py](https://github.com/sparajul/fastmachinelearning/blob/main/preProcessJetImages_santosh.py) -> Prepare Image
- [https://github.com/sparajul/fastmachinelearning/blob/main/prepareTestTrain\\_santosh.py](https://github.com/sparajul/fastmachinelearning/blob/main/prepareTestTrain_santosh.py). -> Prepare test/train
- <https://github.com/sparajul/fastmachinelearning/blob/main/TrainCNN.ipynb> -> Training/hls4ml



# Quark and gluon Images

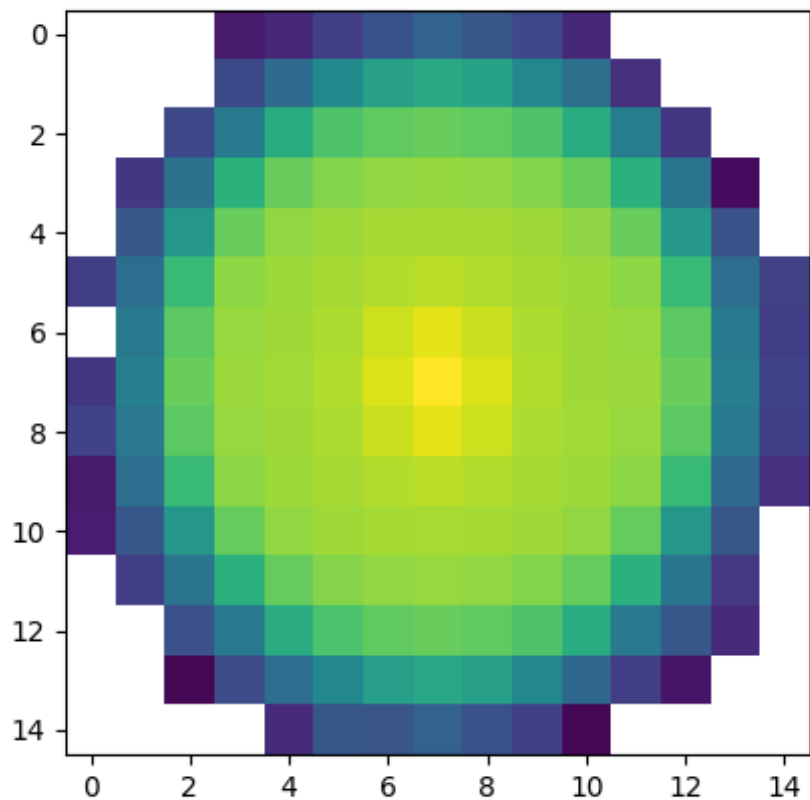


Quark Jets

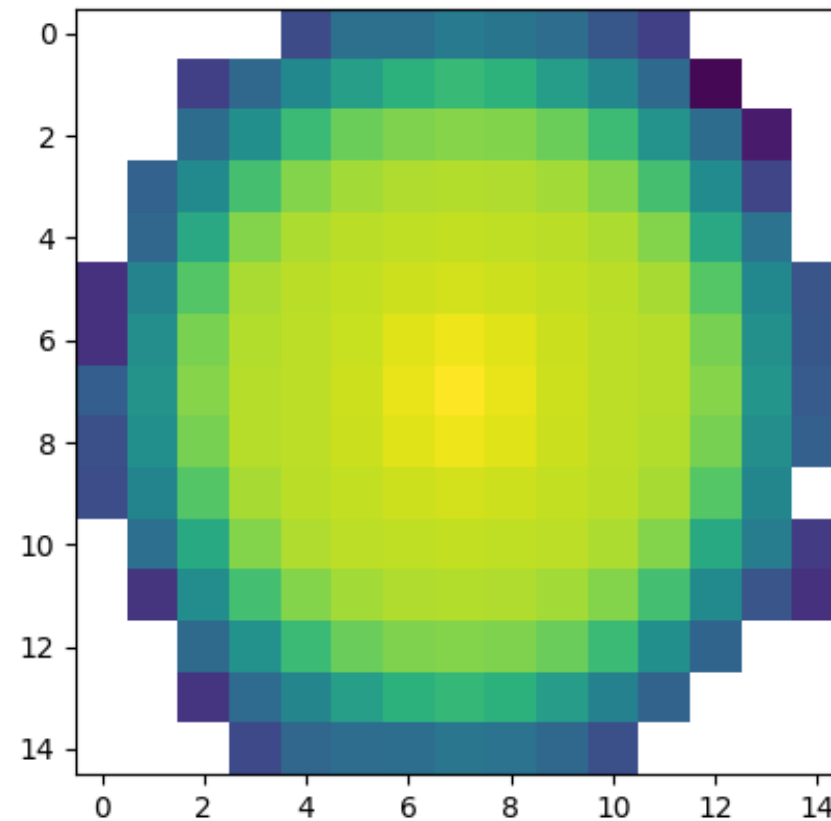
Gluon Jets

Comparison of Quark (left) and Gluon (right) jet images.

## Average Images



Quark Jets



Gluon Jets

Comparison of Quark (left) and Gluon (right) jet images averaged over 60,334 jets.

# CNN Model and Performance:

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten, Dropout, Activation, Conv2D, MaxPooling2D
import tensorflow as tf

model_cnn = Sequential()
model_cnn.add(Conv2D(1, (2, 2), input_shape=(15, 15, 1), activation='relu'))
model_cnn.add(MaxPooling2D(pool_size=(2, 2)))

model_cnn.add(Flatten())
model_cnn.add(Dense(2, activation='softmax'))

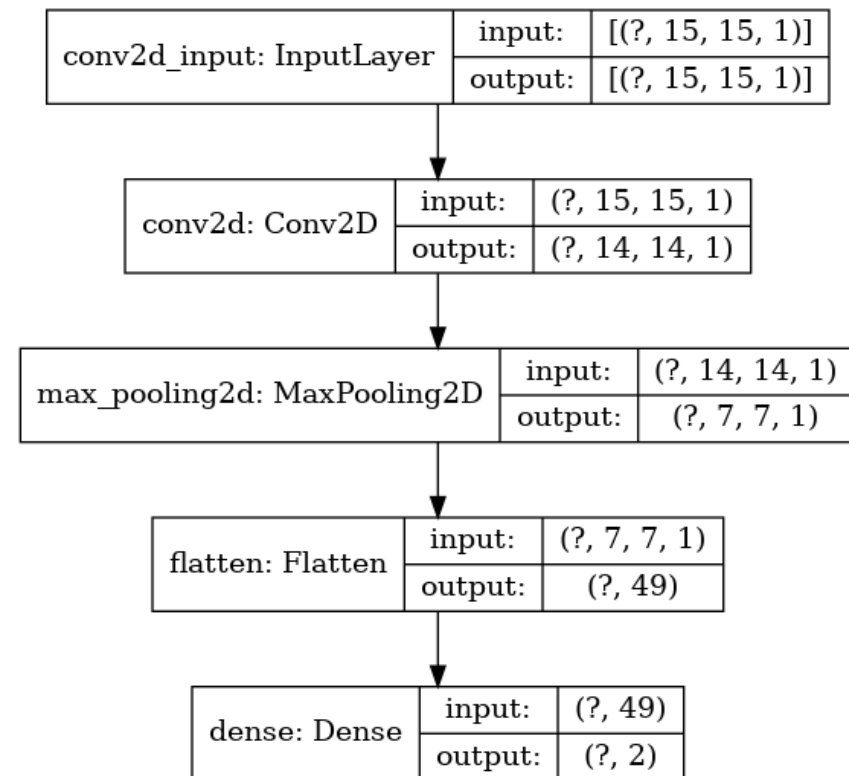
# Compile model

model_cnn.compile(loss='categorical_crossentropy',
                  optimizer=tf.keras.optimizers.Adam(learning_rate=1.0e-4), metrics=['accuracy'])
history_cnn = model_cnn.fit(x_train, y_train, validation_split=0.2, epochs=150,
                           batch_size=256, shuffle=True, verbose=1)
model_cnn.summary()
```

1 Filter

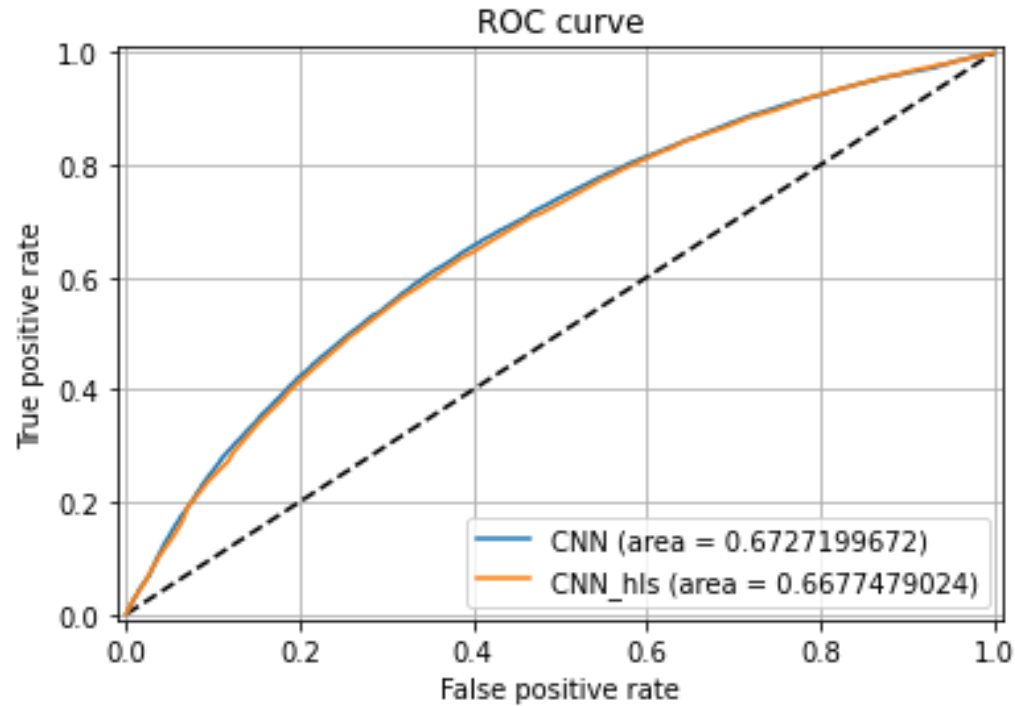
15\*15 grayscale(1) image

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 14, 14, 1)	5
max_pooling2d_1 (MaxPooling2D)	(None, 7, 7, 1)	0
flatten_1 (Flatten)	(None, 49)	0
dense_1 (Dense)	(None, 2)	100
Total params: 105		
Trainable params: 105		
Non-trainable params: 0		



CNN Model

# HIS4ML and Performance:



## SYNTHESIS REPORT:

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== Vivado HLS Report for 'myproject'

=====

\* Date: Wed Mar 8 16:46:27 2023

\* Version: 2018.3 (Build 2405991 on Thu Dec 06 23:56:15 MST 2018)

\* Project: myproject\_prj

\* Solution: solution1

\* Product family: virtexuplus

\* Target device: xcvu13p-flga2577-2L-e

CNN Model/hls4ml Performance

# HIS4ML and Performance:

```
{'EstimatedClockPeriod': '4.156',  
  'BestLatency': '233',  
  'WorstLatency': '233',  
  'IntervalMin': '229',  
  'IntervalMax': '229',  
  'BRAM_18K': '5',  
  'DSP48E': '85',  
  'FF': '2035',  
  'LUT': '4967',  
  'URAM': '0',  
  'AvailableBRAM_18K': '5376',  
  'AvailableDSP48E': '12288',  
  'AvailableFF': '3456000',  
  'AvailableLUT': '1728000',  
  'AvailableURAM': '1280'}
```

```
=====
```

```
== Performance Estimates
```

```
=====
```

```
+ Timing (ns):
```

```
  * Summary:
```

Clock	Target	Estimated	Uncertainty
ap_clk	5.00	4.156	0.62

```
+ Latency (clock cycles):
```

```
  * Summary:
```

Latency		Interval		Pipeline
min	max	min	max	Type
233	233	229	229	dataflow

hls4ml Performance

Latency ~1.2  $\mu$ s

## HIS4ML and Performance:

```
=====
== Utilization Estimates
=====
* Summary:
```

Name	BRAM_18K	DSP48E	FF	LUT	URAM
DSP	-	-	-	-	-
Expression	-	-	0	32	-
FIFO	3	-	122	186	-
Instance	2	85	1907	4713	-
Memory	-	-	-	-	-
Multiplexer	-	-	-	36	-
Register	-	-	6	-	-
Total	5	85	2035	4967	0
Available	5376	12288	3456000	1728000	1280
Utilization (%)	~0	~0	~0	~0	0

```
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```

hls4ml Performance

**THANK YOU!**