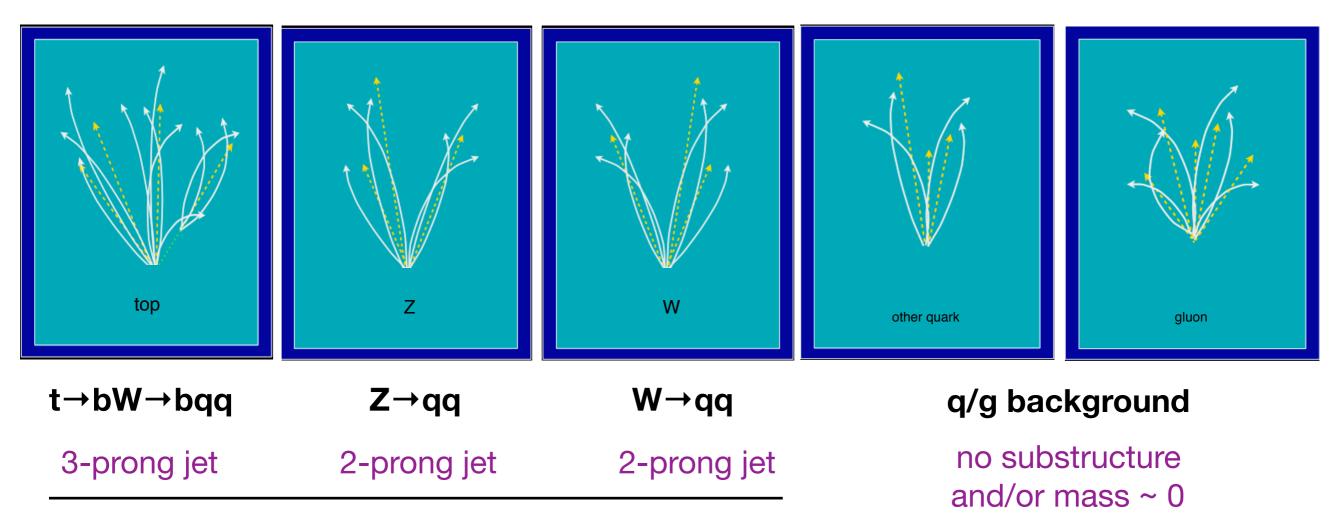
Description of available models

Physics case: jet tagging

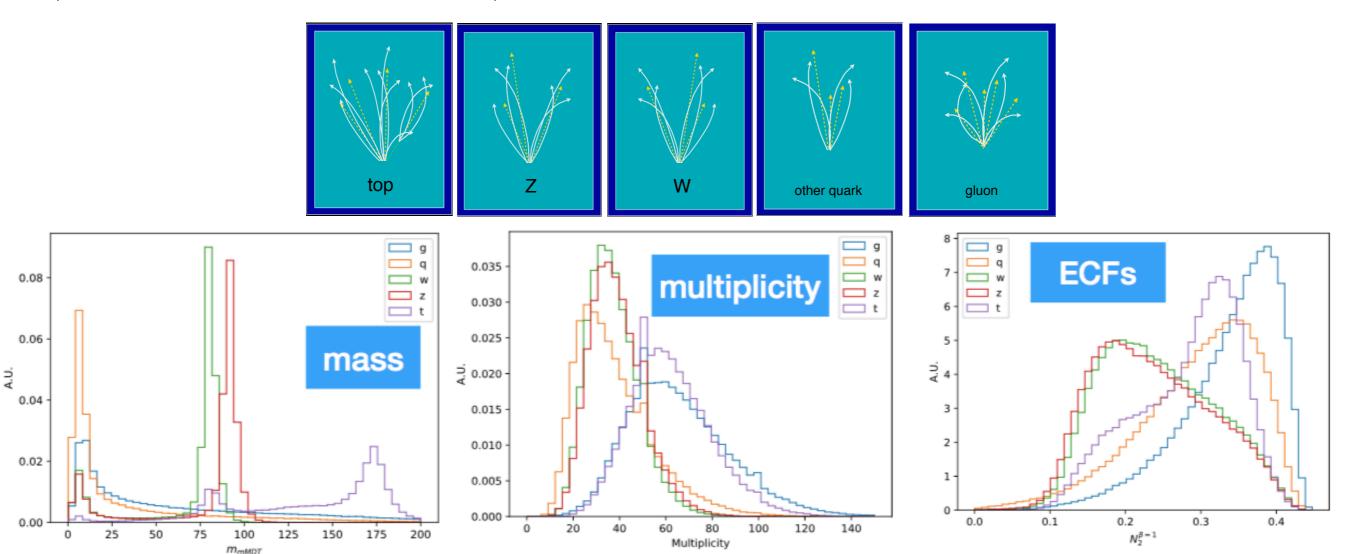
Study a (multi-)classification task to be implemented on FPGA: discrimination between highly energetic (boosted) q, g, W, Z, t initiated jets



Reconstructed as one massive jet with substructure

Physics case: jet tagging

 Dataset: sample of events with a mixture of two boosted WW/ZZ/tt/qq/gg anti-k_T jets (200,000 events for each final state)

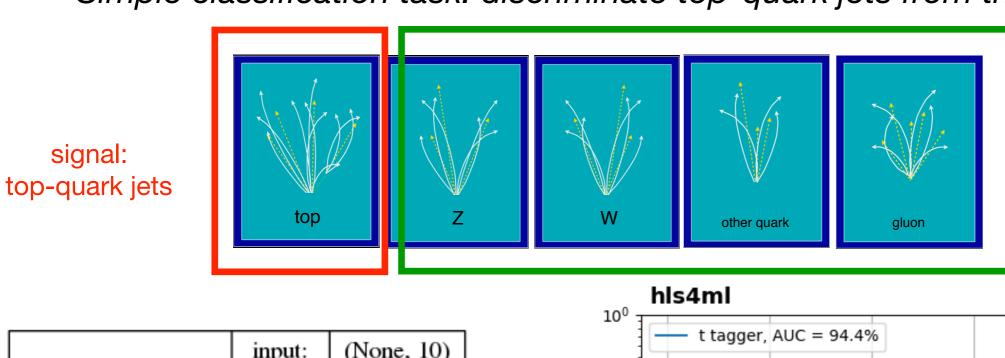


Input variables: several obervables known to have high discrimination power from offline data analyses and published studies [*]

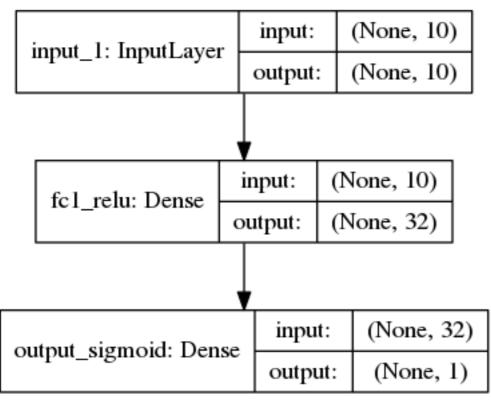
[*] D. Guest at al. PhysRevD.94.112002, G. Kasieczka et al. JHEP05(2017)006, J. M. Butterworth et al. PhysRevLett.100.242001, etc..

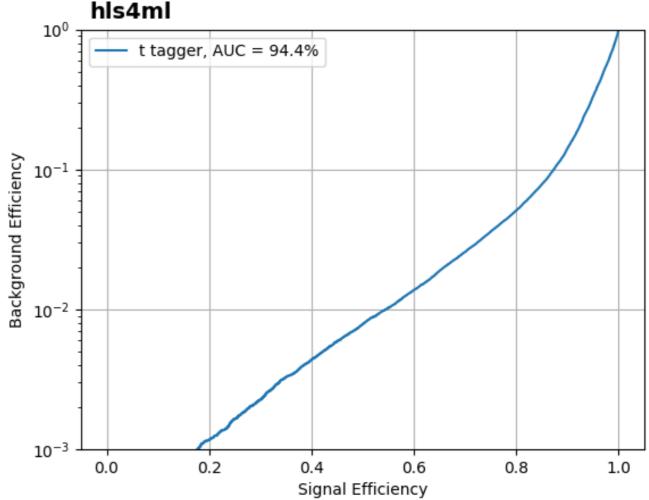
1-layer Dense NN

Simple classification task: discriminate top-quark jets from the others



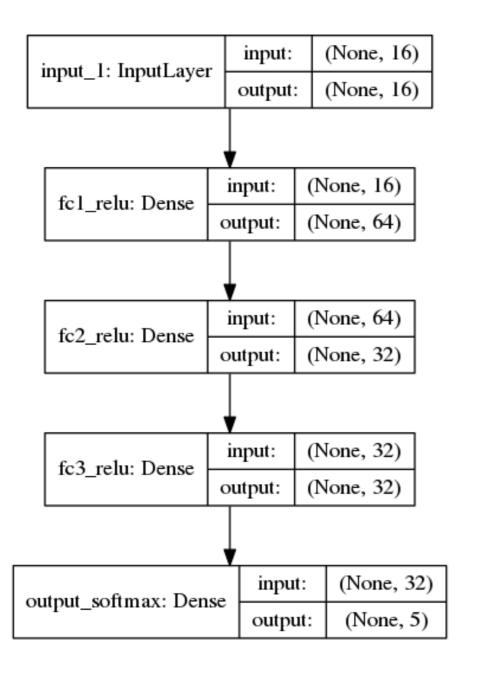
background: all other jets

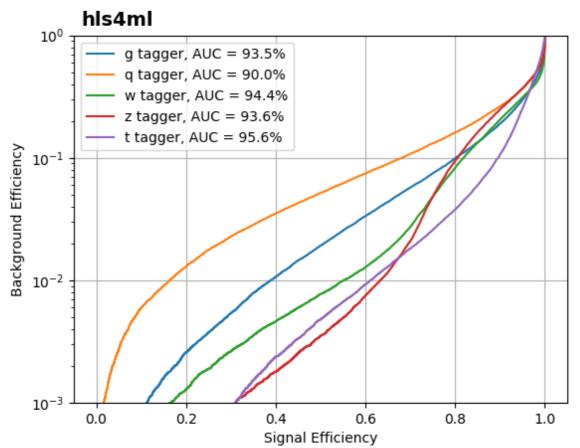


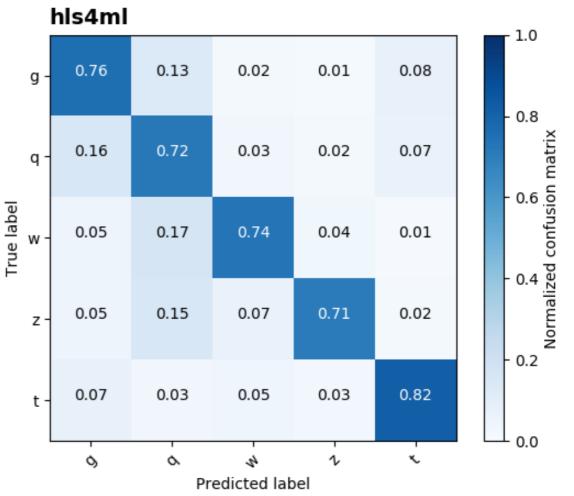


3-layers Dense NN (1)

Multi classification task: discrimination between the 5 classes of jets (t,W,Z,q,g)







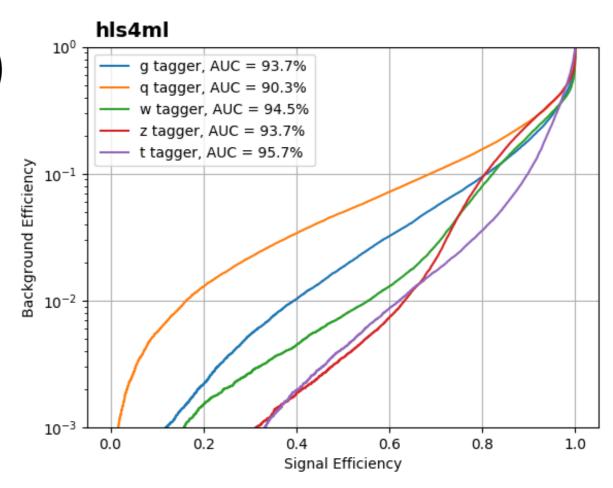
3-layers Dense NN (2)

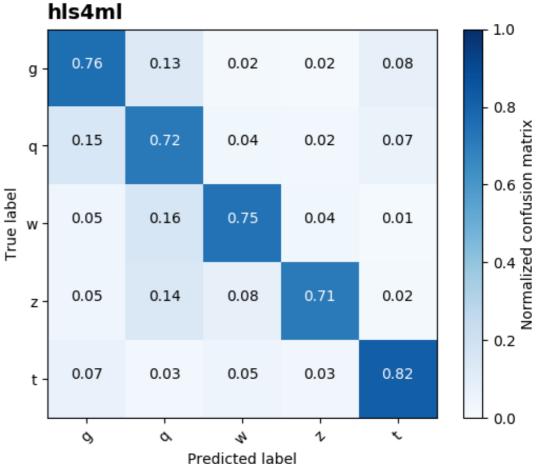
As previous slide but adding batch normalization layers [*]

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	(None, 16)	 0
fc1_relu (Dense)	(None, 64)	1088
bn1 (BatchNormalization)	(None, 64)	256
relu1 (Activation)	(None, 64)	0
fc2_relu (Dense)	(None, 32)	2080
bn2 (BatchNormalization)	(None, 32)	128
relu2 (Activation)	(None, 32)	0
fc3_relu (Dense)	(None, 32)	1056
bn3 (BatchNormalization)	(None, 32)	128
relu3 (Activation)	(None, 32)	0
output_softmax (Dense)	(None, 5)	165
bn4 (BatchNormalization)	(None, 5)	20
softmax (Activation)	(None, 5)	0
	=======================================	==========

Total params: 4,921 Trainable params: 4,655 Non-trainable params: 266

[*] nb, we choose here to do batch normalization before the activation but there are different ways to do this. This results in one more Dense layer (output_softmax).





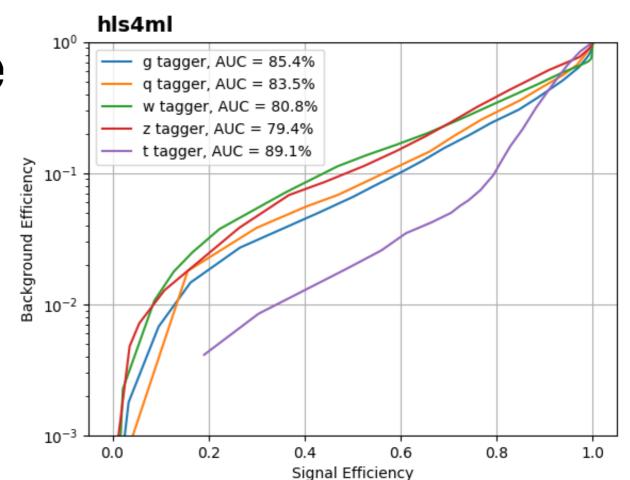
3-layers Binary Dense

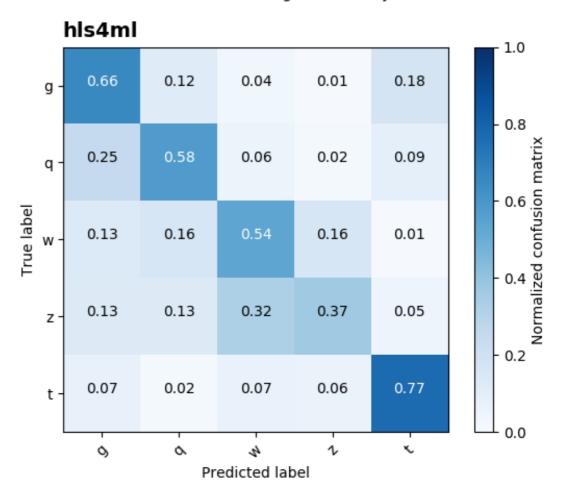
3-layers with batch normalization as in previous slides but:

weights = +/- 1
binary activation function
no output softmax

Layer (type)	Output Shape	Param #
fc1 (BinaryDense)	(None, 64)	1024
bn1 (BatchNormalization)	(None, 64)	256
act1 (Activation)	(None, 64)	0
fc2 (BinaryDense)	(None, 32)	2048
bn2 (BatchNormalization)	(None, 32)	128
act2 (Activation)	(None, 32)	0
fc3 (BinaryDense)	(None, 32)	1024
bn3 (BatchNormalization)	(None, 32)	128
act3 (Activation)	(None, 32)	0
output (BinaryDense)	(None, 5)	160
bn (BatchNormalization)	(None, 5)	20
Total params: 4,788		

Trainable params: 4,788
Non-trainable params: 266





3-layers Ternary Dense

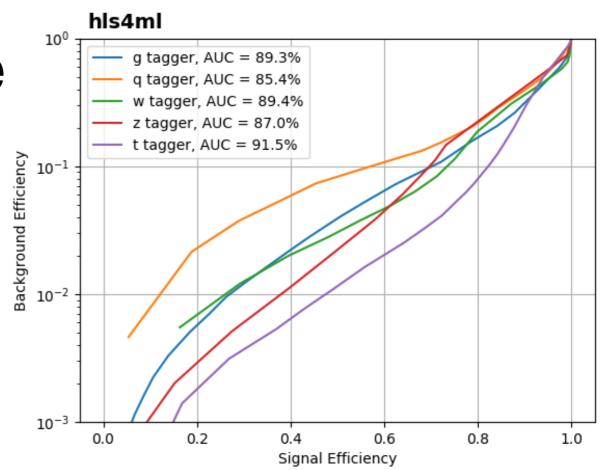
3-layers with batch normalization as in previous slides but:

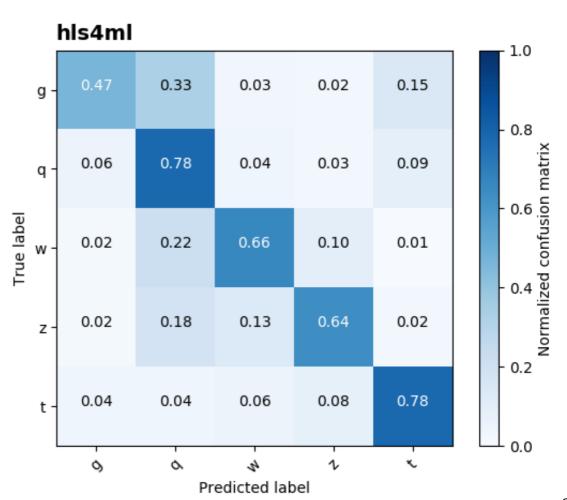
weights = +/- 1 or 0
ternary activation function
no output softmax

Layer (type)	Output Shape	Param #
fc1 (TernaryDense)	(None, 64)	1024
bn1 (BatchNormalization)	(None, 64)	256
act1 (Activation)	(None, 64)	0
fc2 (TernaryDense)	(None, 32)	2048
bn2 (BatchNormalization)	(None, 32)	128
act2 (Activation)	(None, 32)	0
fc3 (TernaryDense)	(None, 32)	1024
bn3 (BatchNormalization)	(None, 32)	128
act3 (Activation)	(None, 32)	0
output (TernaryDense)	(None, 5)	160
bn (BatchNormalization)	(None, 5)	20
Total params: 4,788		===================================

Total params: 4,788
Trainable params: 4,522
Non-trainable params: 266

[*] more info: https://arxiv.org/abs/1605.04711





List of available models

- Find the weigths (.h5 file) and model architecture (.json file) under the example-kerasmodel-files
 - 1-layer Dense NN:
 KERAS_1layer.json, KERAS_1layer_weights.h5
 - **3-layers Dense NN:**KERAS_3layer.json, KERAS_3layer_weights.h5
 - 3-layers Dense NN + BatchNormalization:
 KERAS_3layer_batch_norm.json, KERAS_3layer_batch_norm_weights.h5
 - **3-layers Binary Dense NN:** KERAS_3layer_binary_weights.h5
 - **3-layers Ternary Dense NN:**KERAS_3layer_ternary.json, KERAS_3layer_ternary_weights.h5

Pick your favourite one to implement it on FPGA!