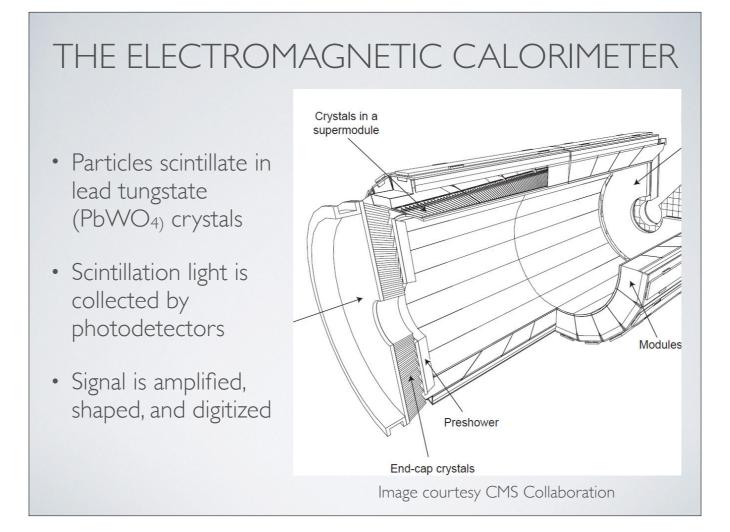
# STUDYING THE CMS ECAL TRIGGER AT THE CERN LHC

Kirsten Randle Prof. Toyoko Orimoto

#### THE ELECTROMAGNETIC CALORIMETER

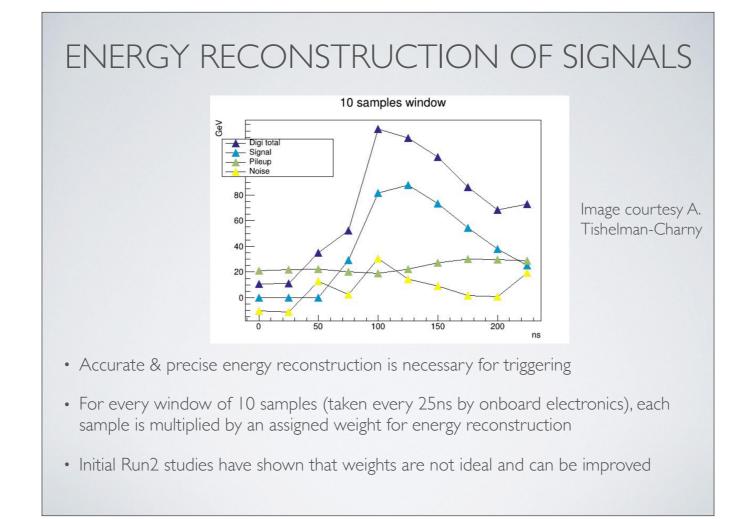
- The Compact Muon Solenoid (CMS) detects proton-proton collisions at the CERN LHC
- CMS is made up of several layers that study different properties of particles
- The Electromagnetic Calorimeter (ECAL) is the layer that is sensitive to photons and electrons.

layers: -silicon tracker, -e cal, -hcal, -solenoid (not a detector, produces B-field) , - muon chambers



CMS is split into 2 parts -barrel -endcaps

Crystals are first in groups of 5 crystals for one board (VFE -very front-end) and then in larger groups that depend on the angle from the beam-axis (  $eta = -ln(tan(\theta)/2)$  )



Weights are designed to preferentially amplify the peak amplitude over the background to get more accurate energy reconstruction of the event

## WEIGHTS

- Amplitude weights
  - · active, uniform for whole detector
  - Current study: update weights and vary weights for different parts of the detector
- Timing weights
  - VFE boards have unused capacity for second set of weights
  - Current study: optimize timing weights to identify out-of-time pileup in a signal

Reminder: VFE -very front end

Crystals in high-eta regions, like the endcaps are especially susceptible to radiation damage and may need unique weights, while weights are currently uniform for the whole detector

## PILEUP

- Bunches of protons meet inside the detector every 25 ns, this is called a Bunch Crossing (BX)
- Scintillation in the detector takes much longer than this (10 samples, 250 ns)
- Pieces of signals from other bunch crossings can add to overall amplitude
- Out-of-time pileup not only changes the amplitude but also the pulse shape

# SCOPE OF PROJECT

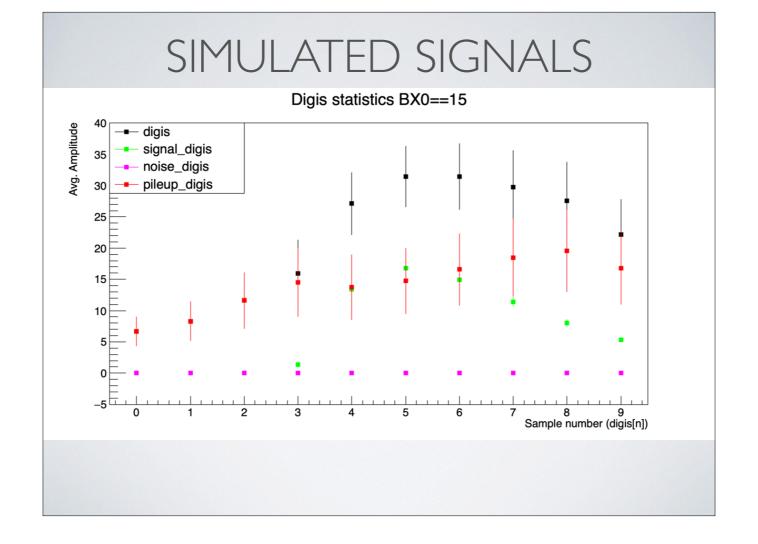
- Get ROOT and pyROOT running
- Write a flexible plotter that uses ROOT
- Produce interesting of plots of various parameters used to study the amplitude and timing weights from simulated data

ROOT is C++ based particle physics analysis framework pyROOT — root libraries can be imported into python do two layers of code: first use a layer

## **PLOTTER**

- Plotter is split into 2 layers to separate time-heavy processes for ease of use
- ROOT libraries are used for power to deal with large datasets
- First layer takes data and produces histogram objects and saves them to a file that can be accessed for the second layer
- Second layer takes histogram objects and plots them, since second layer runs quickly it can be tweaked easily
- Flexibility makes changes to cuts or repeating the same studies on different data sets or changing cuts

```
65 def create_1Dhisto(bias_tree, histo_name, binparams, parameter, cuts):
                h = TH1F(histo_name,parameter,binparams[0],binparams[1],binparams[2])
                 h.GetXaxis().SetTitle(parameter)
67
68
                 h.GetYaxis().SetTitle('Entries')
                  drawstatement = parameter + ' >> ' + histo_name
69
70
                  bias_tree.Draw(drawstatement,cuts,'hist')
71
                  h.SetDirectory(0)
72
                  return h
73
74 def create_2Dhisto(bias_tree,histo_name,binparams,parameters,cuts):
75
                             TH2F (histo\_name, histo\_name, binparams[0][0], binparams[0][1], binparams[0][2], binparams[1][0], binparam
                             [1][1],binparams[1][2])
76
                  h.GetXaxis().SetTitle(parameters[0])
                  h.GetYaxis().SetTitle(parameters[1])
77
                  drawstatement= parameters[1] + ':' + parameters[0] + ' >> ' + histo_name
78
                  bias_tree.Draw(drawstatement,cuts,'COLZ1')
79
80
                  h.SetDirectory(0)
81
                  return h
82
83 def slicefity(histo,func,slicebins,options):
                   fitparams = TObjArray()
                  histo.FitSlicesY(func,slicebins[0],slicebins[1],slicebins[2],options,fitparams)
                  return fitparams
86
87
88 def iterate_curves(tree,curvelist,type):
                  list = [0 for x in range(len(curvelist))]
89
90
                  for i,p in enumerate(curvelist):
91
                           if type == 'TH1F':
                                     list[i] = create_1Dhisto(tree,p[0],p[2],p[3],p[4])
92
                            if type == 'TH2F':
93
94
                                     list[i] = create_2Dhisto(tree,p[0],p[2],p[3],p[4])
95
                 return list
```



The average signal (green)

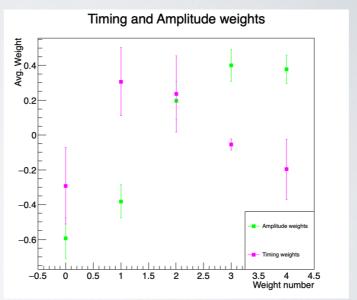
Average electronic noise (magenta) (can be positive or negative, small amplitude, average is zero)

Pileup (red) is very variable— only 1 BX is studied, which means that different bunch crossings may have different pileup behavior (BXs with more or fewer events preceding/following it will have different distributions)

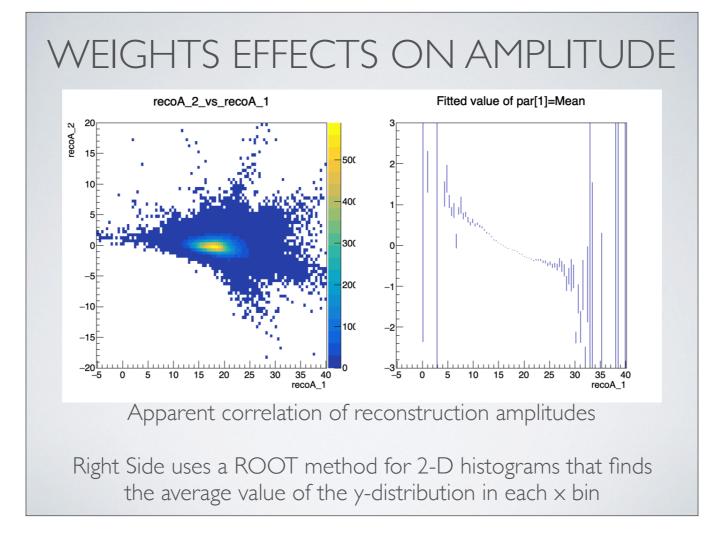
This plot is at eta ring 28, bunch crossing 15

## WEIGHTS

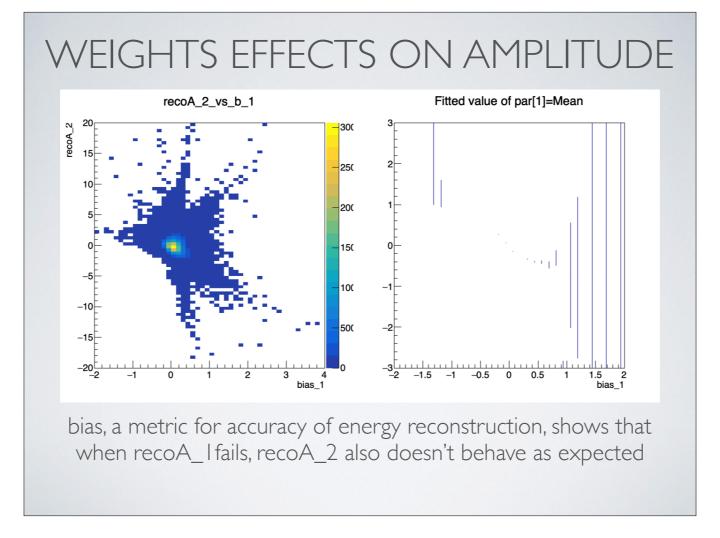
- The model uses the 'true' amplitude to choose the weights that accurately reconstruct the amplitude for each event
- The model similarly chooses the timing weights that reconstruct the amplitude to zero for each event



Check with Abe on how accurate description of timing weights is



Examining behavior of recoA\_2 vs recoA\_1: we expect the average of recoA\_2 to be constant at zero, and for recoA\_1 and recoA\_2 to be uncorrelated



Examining behavior of recoA\_2 vs recoA\_1: we expect the average of recoA\_2 to be constant at zero, and for recoA\_1 and recoA\_2 to be uncorrelated

## FUTURE DIRECTIONS

- Determine what types of events are failing to produce appropriate timing weights (recoA are far from expected values)
- Quantify in which sample high pileup affects amplitude reconstruction the most
- Study failure modes for amplitude weight failure, specifically the pileup dependence
- Examine Timing weight's ability to identify out-of-time pileup

Determine which events are failing — this may require examining single events  ${\sf Quantify}$