Advanced Classification Techniques

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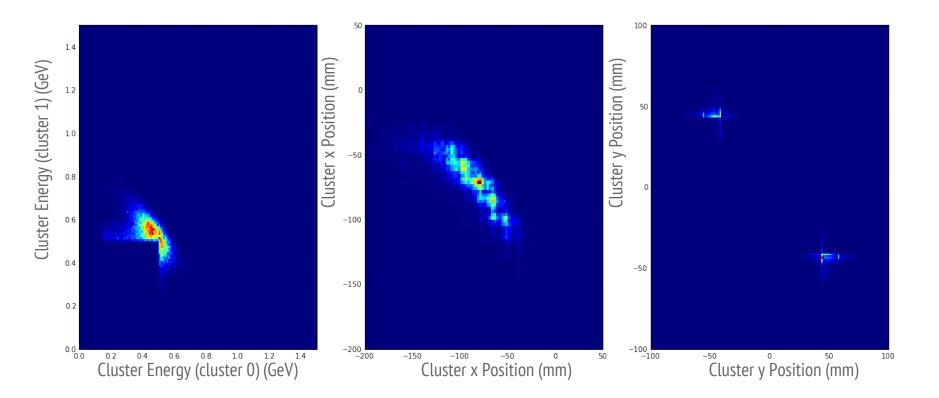
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Heavy Photon Search Collaboration Meeting October 26-28, 2015

Preliminaries

- Wanted to understand if classification of events using any one of the machine learning algorithms is more efficient than using a manual selection
- Started using TMVA but switched to scikit-learn
- As a proof of concept, begin by trying to classify Mollers using a simple decision tree and ensemble methods
- Train and validate all algorithms using the pass 3 MC Moller sample as the signal and MC beam-tri files, with some cuts, as the background
- Run over run 5772 in order to check that the algorithms are reasonable
- All files (MC and data) are preprocessed
 - Only look at events that have two clusters in opposite volumes and are within a time window of 1.6 ns
 - Require that both clusters have a track matched to them
- A flat ntuple containing the following variables is made out of the events that pass the cut:
 - Energies and position of Ecal cluster pair
 - Track momentum, parameters and charge of the tracks matched to the clusters

Signal Preprocessing



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Signal Track Variables

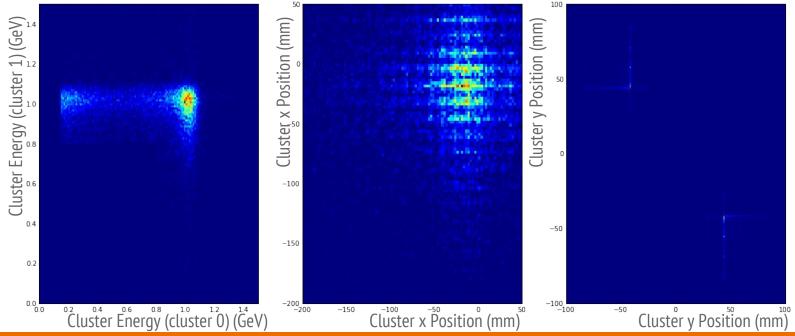
green: track 1 b (GeV) 1000 800 0.6 600 0.4 400 0.2 200 0.0 -0.06 p (GeV)¹⁰ -0.04 -0.02 0.2 0.8 0.00 0.02 0.04 p_x (GeV) 0.4 0.6 1800 1000 1600 800 1400 1200 600 1000 800 400 600 400 200 200 p_y (GeV)^{0.06} -0.06 0.0 0.04 0.2 -0.04 -0.02 0.00 0.02 0.4 0.6 0.8 1.0 p_z (GeV)

blue: track 0 - track associated with highest energy clusters

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Background Preprocessing

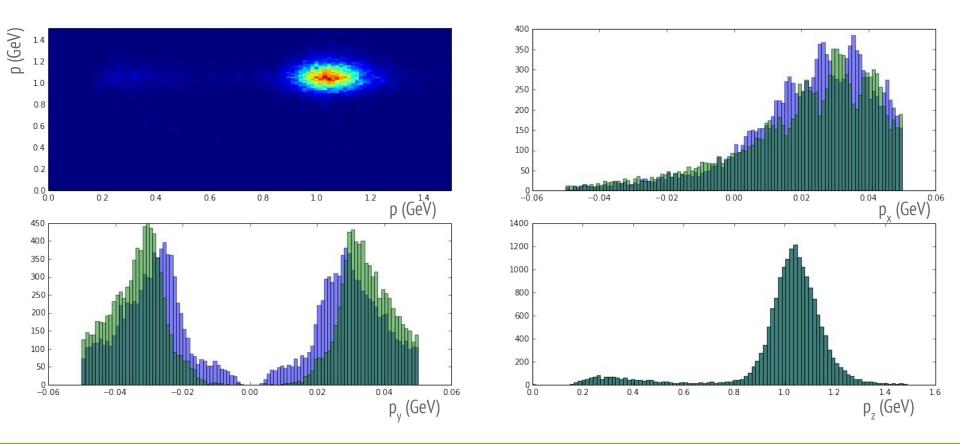
- Apply same requirements as signal but ...
- Don't have background files without Mollers so require that one of the clusters in the pair have an energy > .8 GeV cut to remove possible moller candidates



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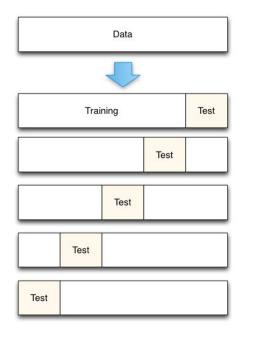
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Background Track Variables



Cross Validation

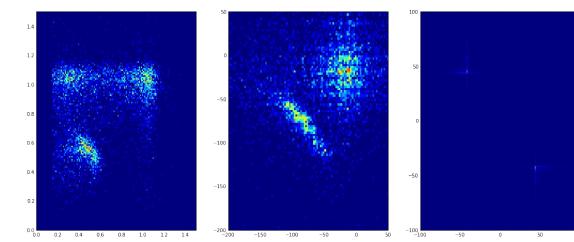
- Combine signal and background sample and then break them up into 20 equal samples
- Train the algorithms using 19 out of the 20 samples
- Calculate the efficiency to separate signal from background and their means as their score

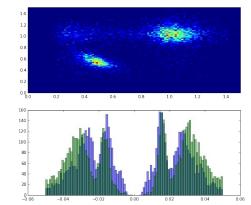


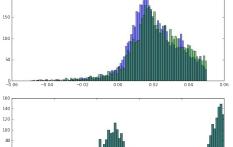
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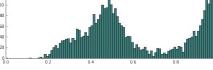
Classifier	Score
Decision Tree	0.999410359802
Random Forest	0.99929246999770327

Using Real Data







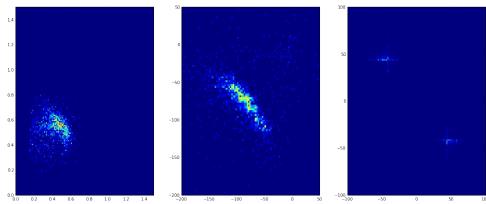


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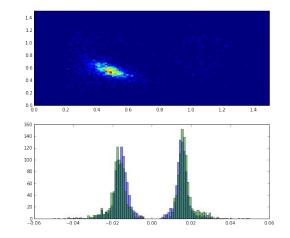
Decision Tree

-200

1.4

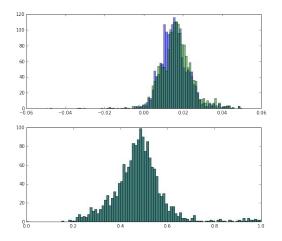


-100



100

-50



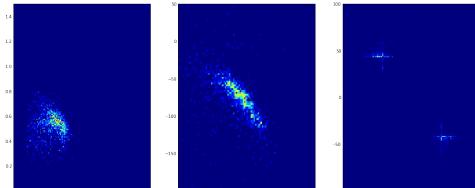
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0.0

0.4

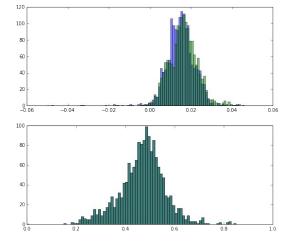
0.8

Random Forest



0.0 0.2 0.6 0.8 1.4 -200 -100 -100 -100 50

1.4 1.2 1.0 0.8 0.6 0.4 0.2 0.0 **•** 0.0 1.2 1.4 0.2 0.4 0.6 0.8 1.0 160 140 120 100 -0.06 -0.04 -0.02 0.00 0.04 0.06



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