

# THE ‘MULTIPLICIY JUMP!’ – FINDING B’S IN MULTI-TEV JETS W/O TRACKS

**TODD HUFFMAN, OXFORD UNIVERSITY**

**THOMAS RUSSELL, CURRENTLY @ BLOOMBERG L.P.  
JEFF TSENG, OXFORD UNIVERSITY**

**[ARXIV:1701.06832](https://arxiv.org/abs/1701.06832)**

**[ALSO: 2016 J.PHYS. G: NUCL. PART. PHYS. 43 085001](#)**

# OUTLINE

Why do we want to tag b-jets?

Explain the “multiplicity jump” technique  
(aka – Todd’s crazy idea)

The “invented” parameter  $F_i$

Pile-up and luminous region

Using an Artificial Neural Net (ANN)

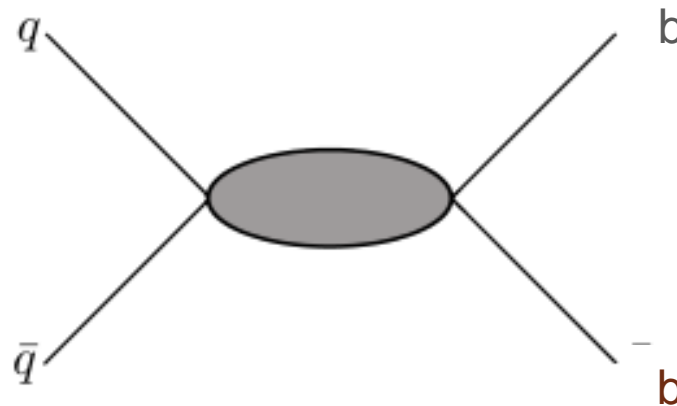
- (It’s not “deep” enough to be a DNN)
- Learning a little from “deep learning”

Future prospects @ ATLAS

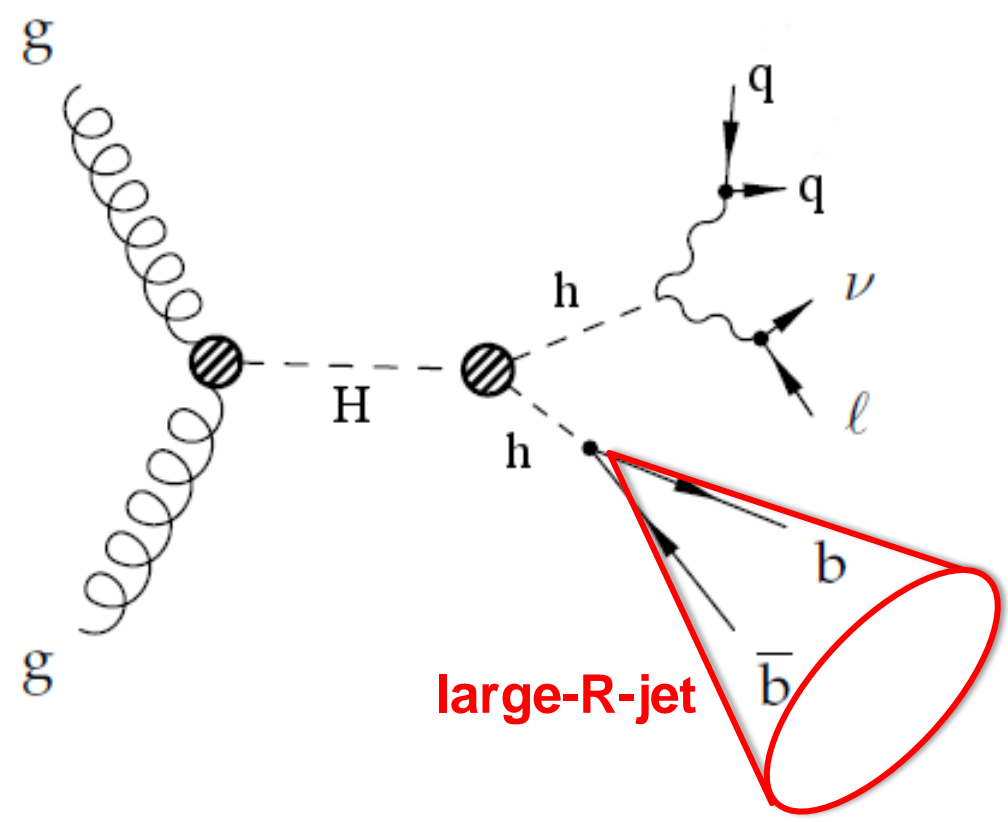
# B HADRON'S IN JETS? WHY BOTHER -

Goal: Search for non-resonant new physics using angular distribution of dijet events

e.g. Contact Interaction (CI)



**1-lepton final state**  
 $hh \rightarrow bb + WW^* \rightarrow bb + lvqq$



**Boosted analysis**

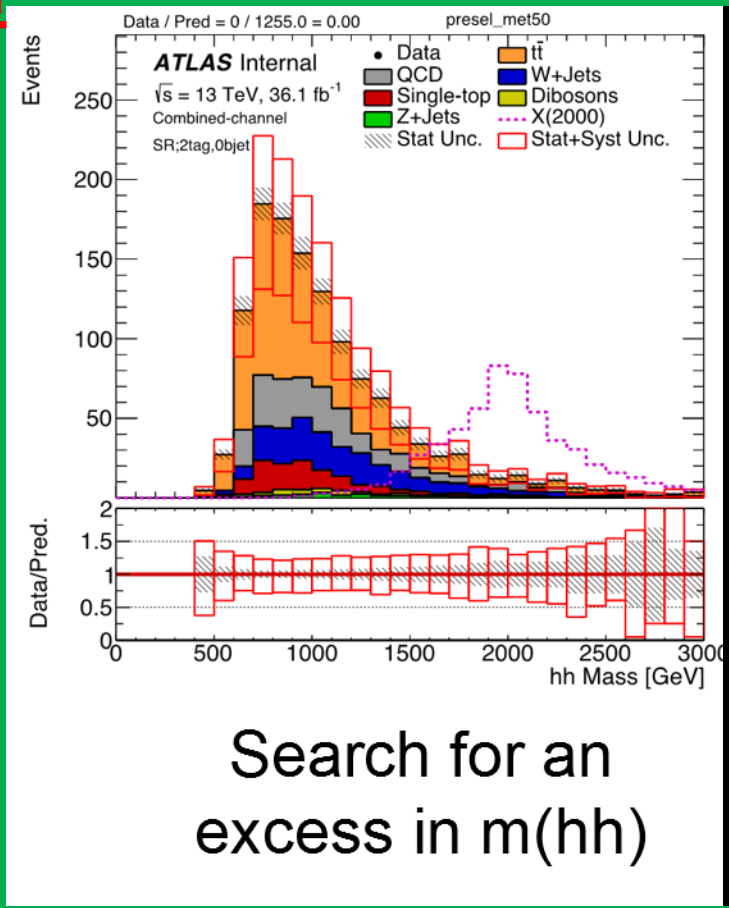
$h \rightarrow bb$  reconstruction with large-R jet.

# BASIC POINT OF SEARCHES

Falling, smooth background

New Physics

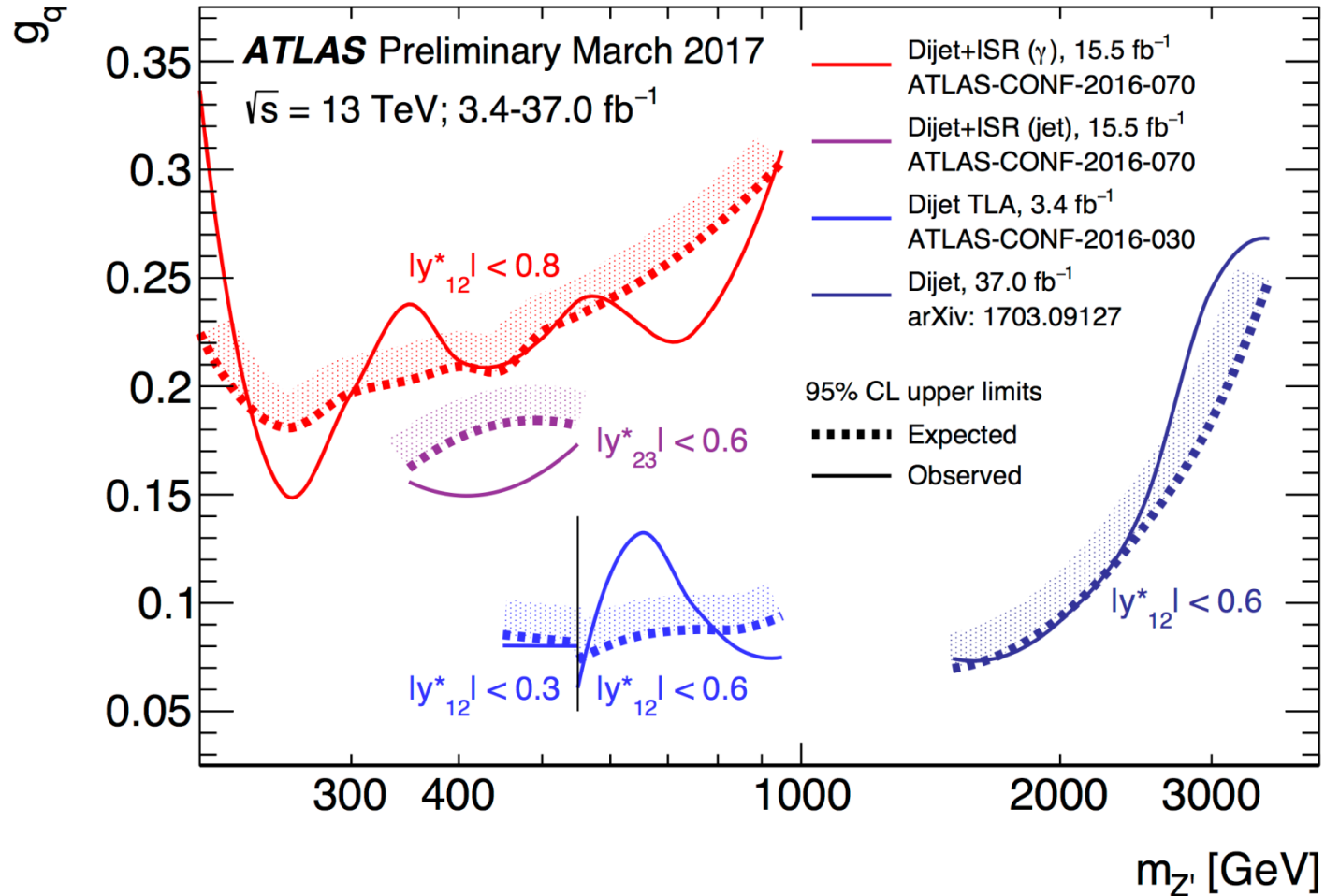
Invariant Mass

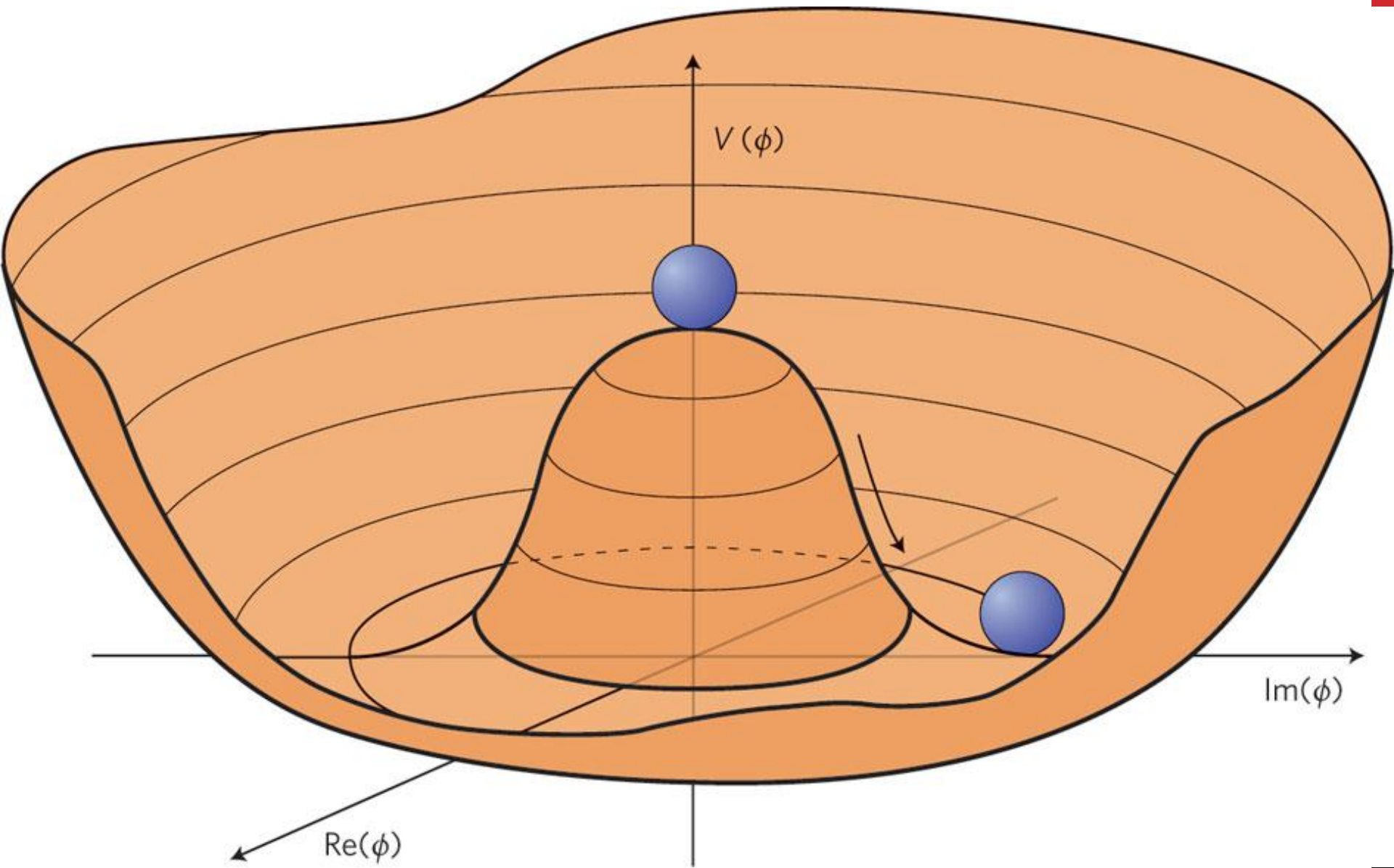


# DIJETS

## ATLAS SEARCH - 2016

want to b-tag these jets

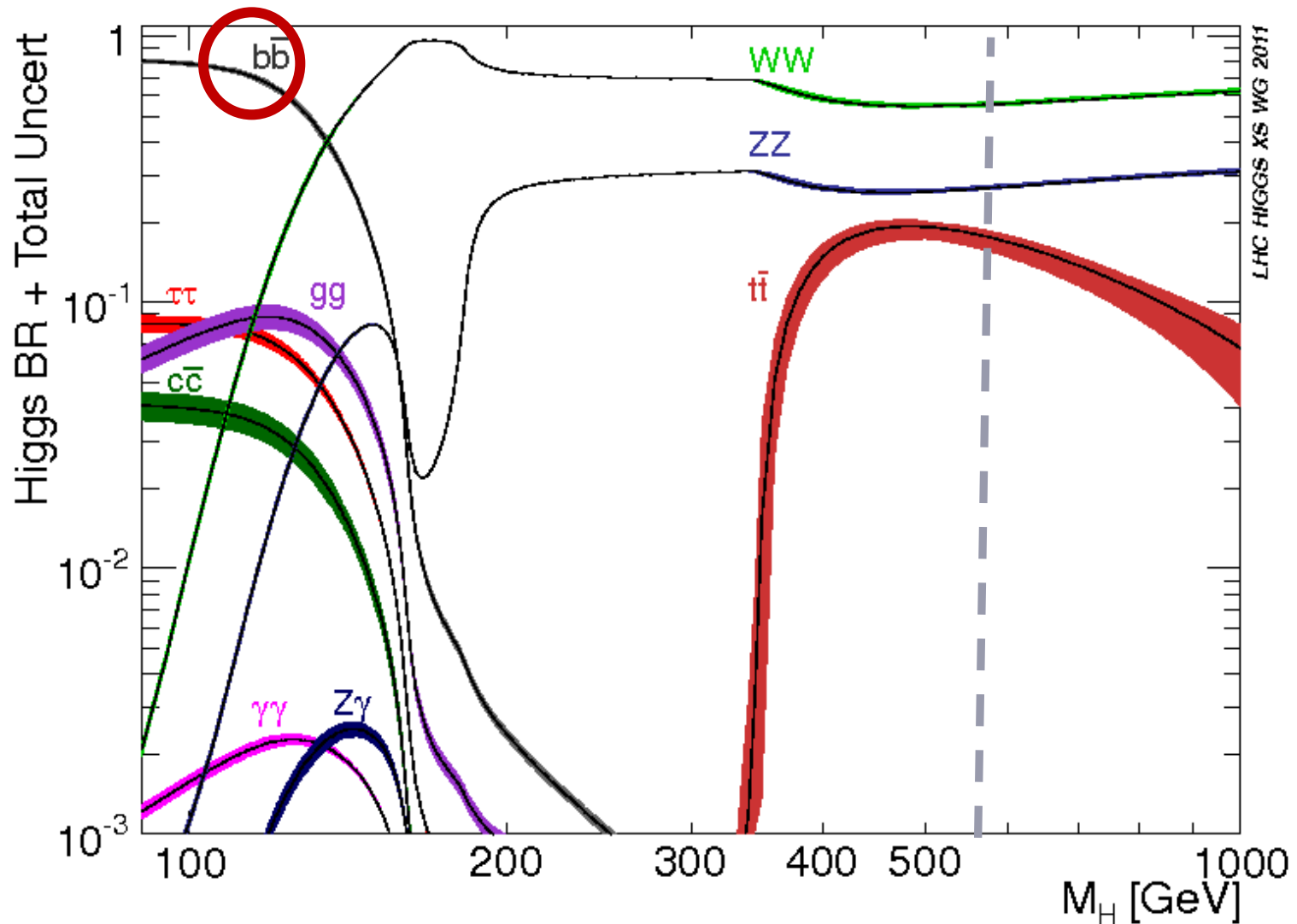




Taken from *Nature Physics* **7**, 2–3 (2011);

# $H \rightarrow BB$

# 80% BRANCHING RATIO

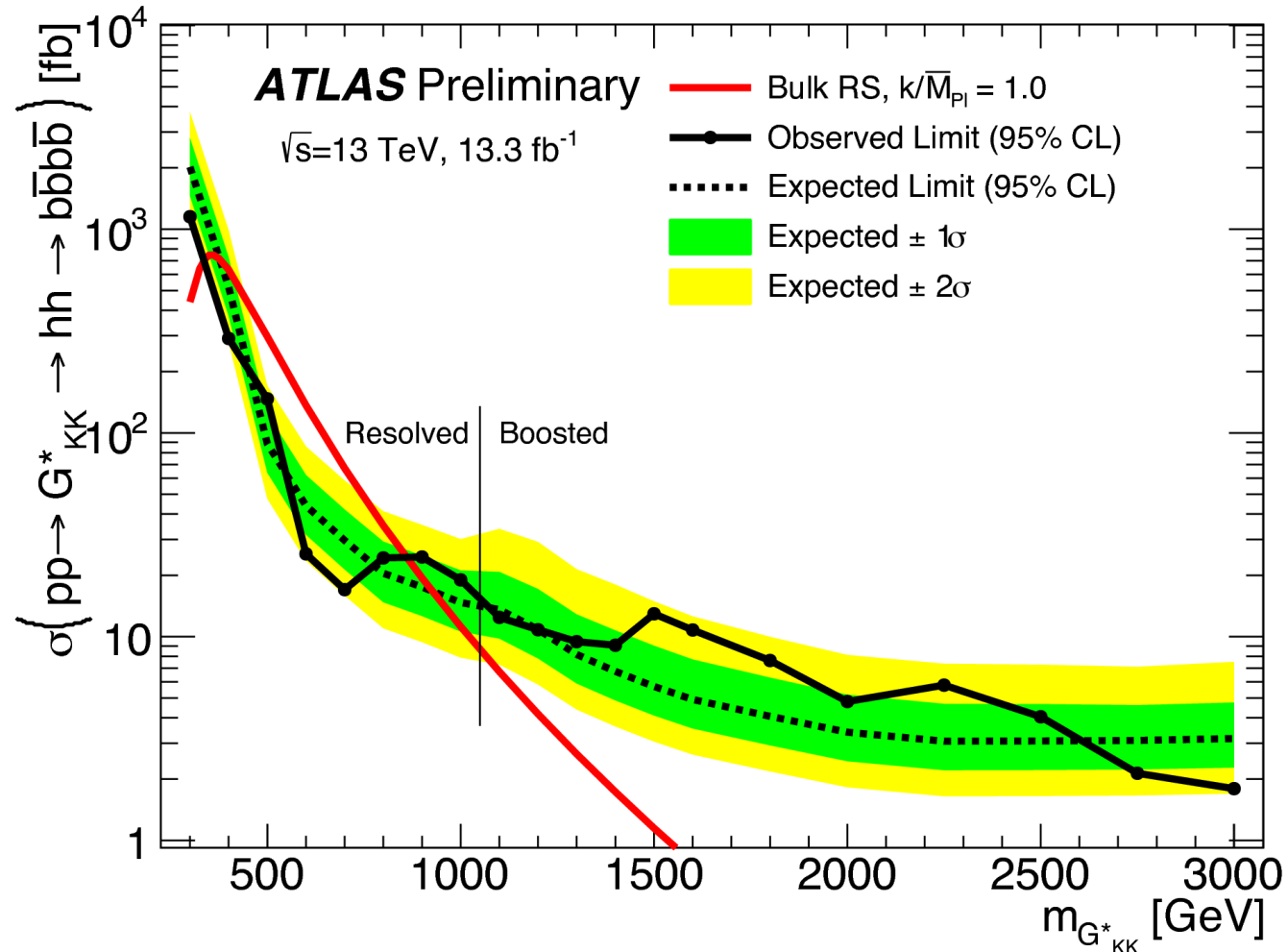




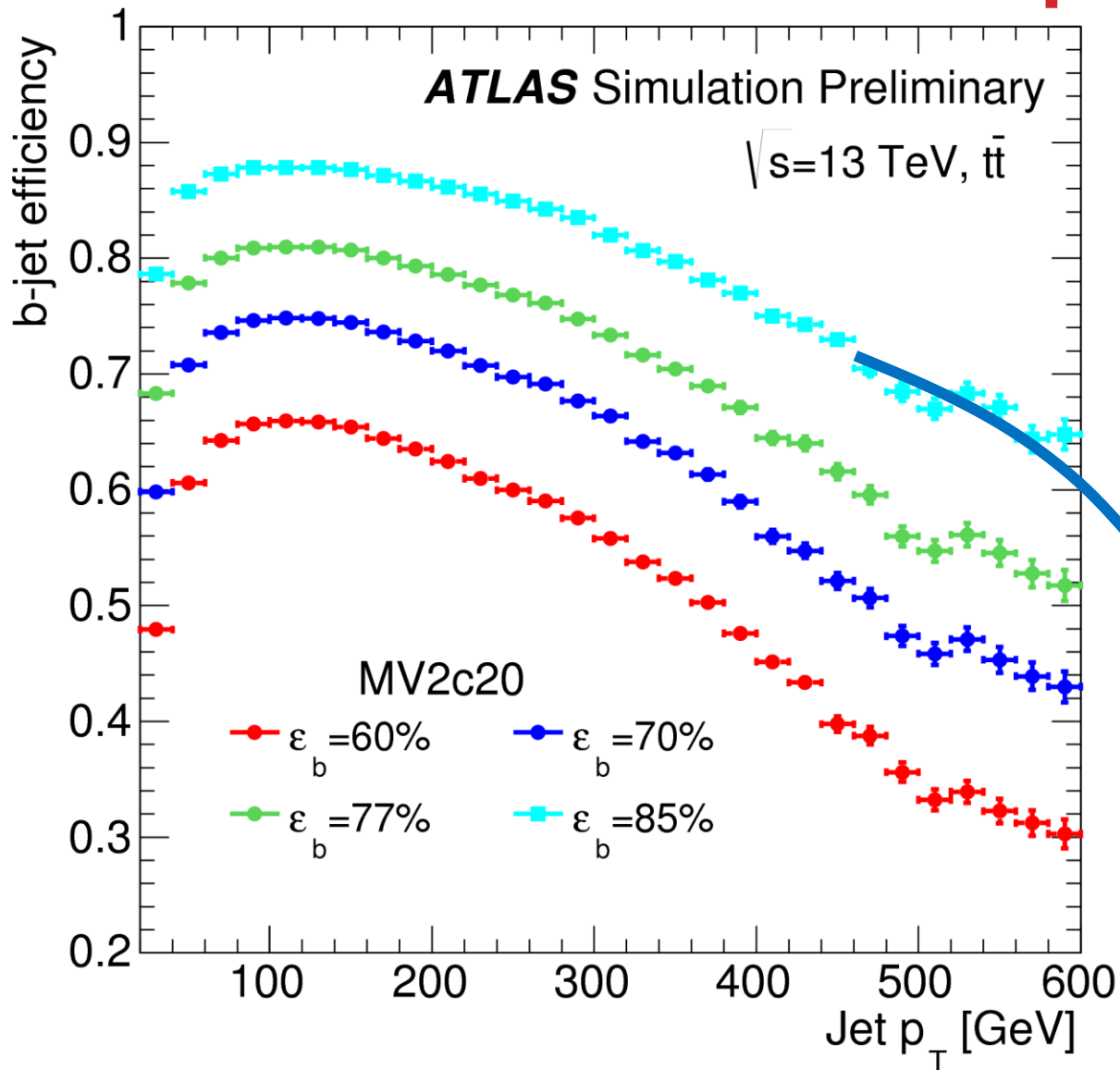
# HIGGS HUNTING EXAMPLE

## HH $\rightarrow$ 4B'S

ATLAS-CONF-2016-049



# PROBLEM: TAG HIGH $p_T$ B-JET



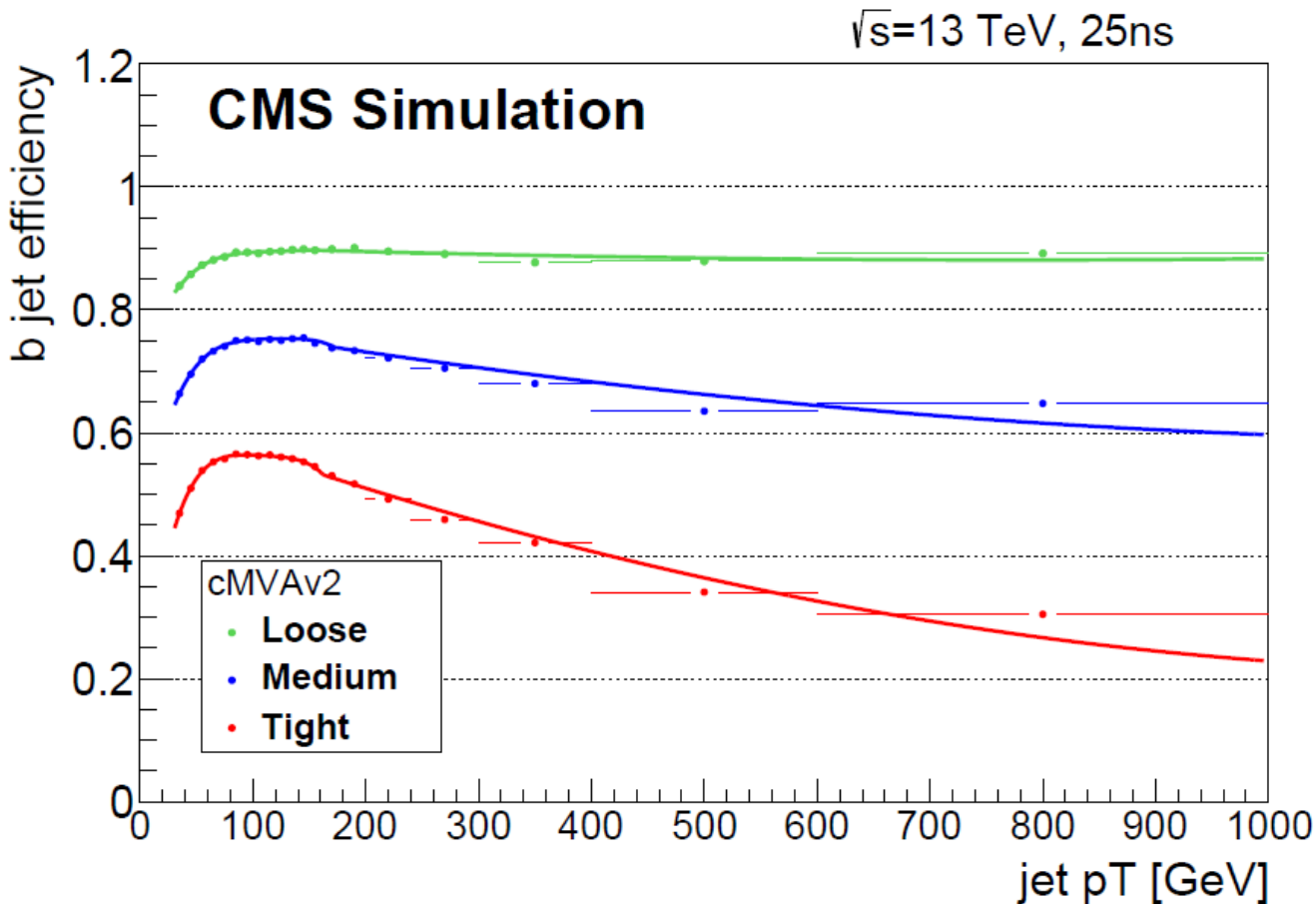
Aux. Fig. 4a  $\rightarrow$  Pub. Note  
[ATL-PHYS-PUB-2015-022](#)

Next: Why it happens

# PROBLEM: TAG HIGH $P_T$ B-JET

Much great work happening

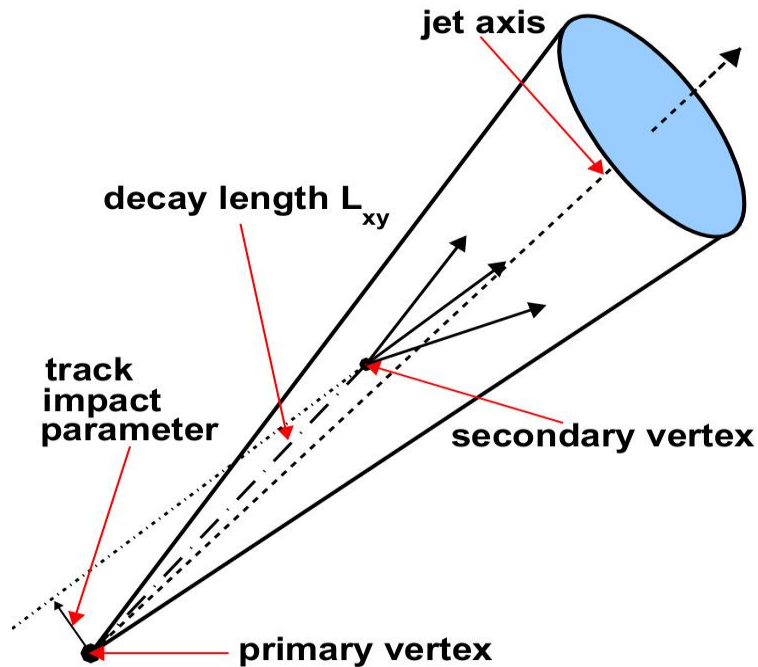
From CMS btag note  
[CMS PAS BTV-15-001](#)



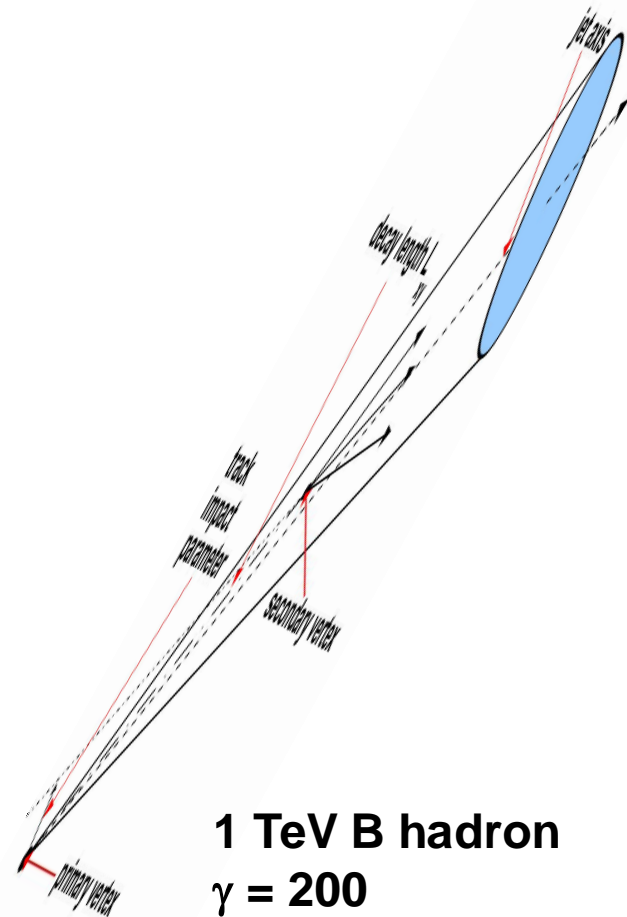
# B-TAGGERS

## FIGHT RELATIVITY

Small cone sizes are prevalent.  
 $\Delta R \approx 0.04$  for a B in a 500+ GeV Jet.

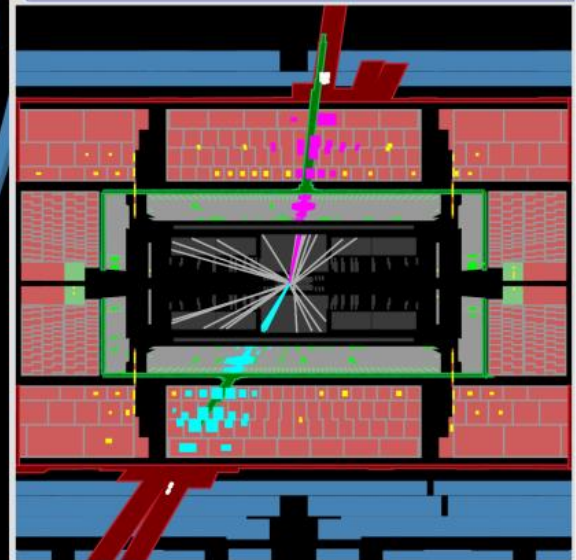
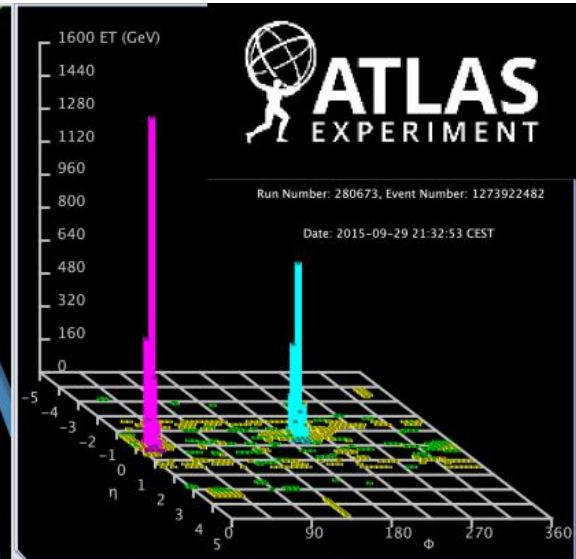
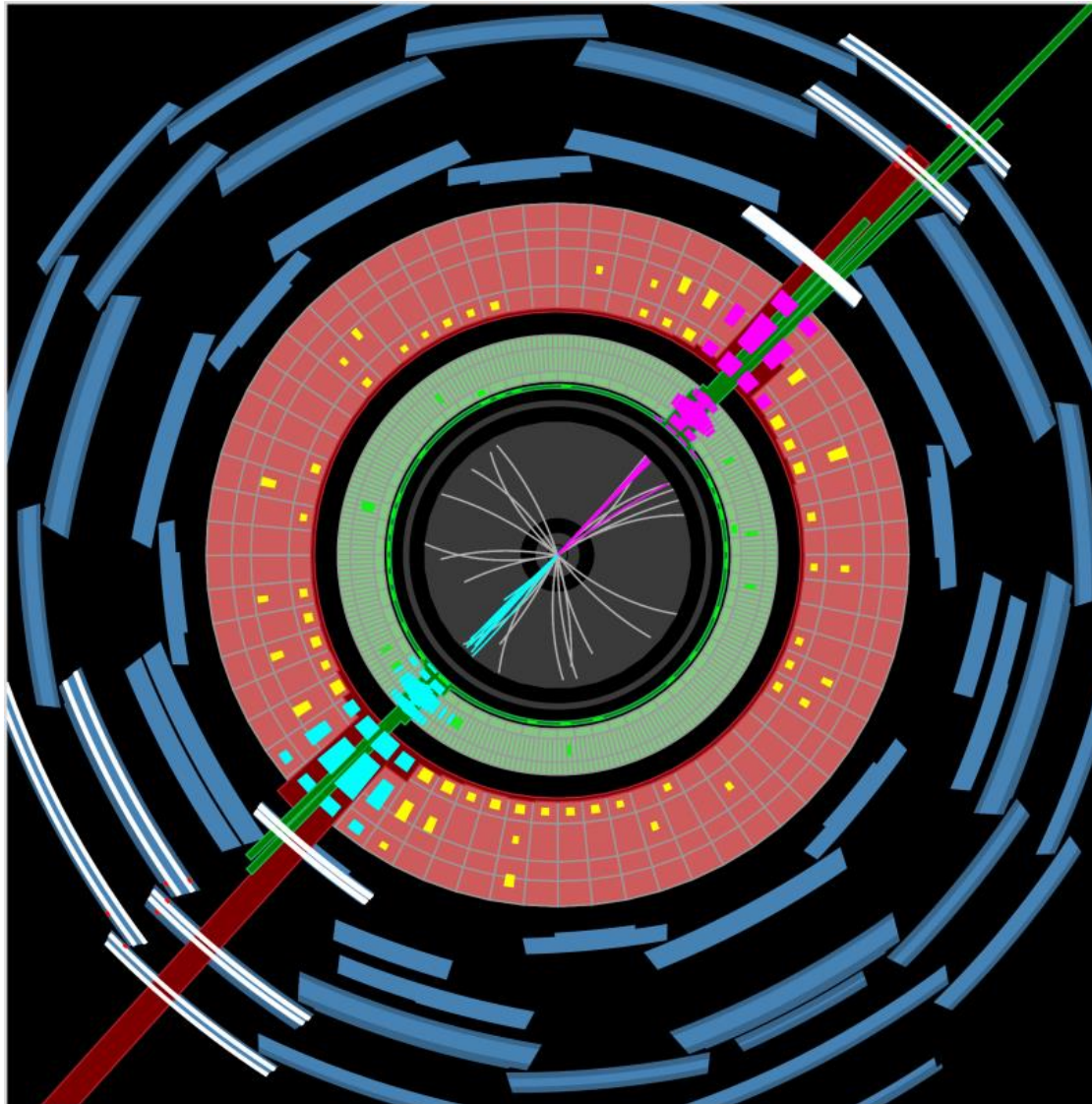


**200 GeV B hadron**  
 $\gamma = 40$   
 $\gamma c\tau = 18 \text{ mm}$

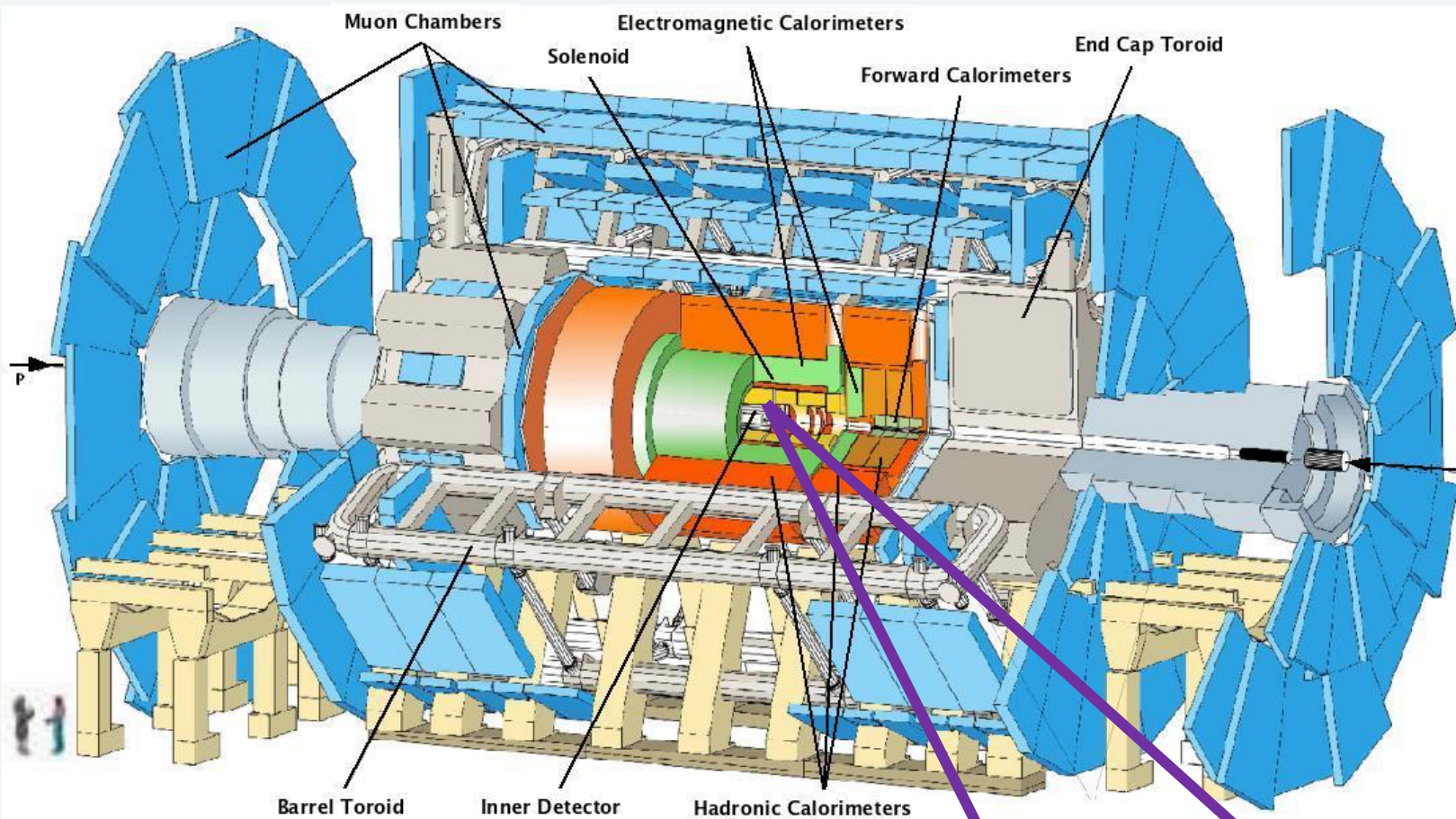


**1 TeV B hadron**  
 $\gamma = 200$   
 $\gamma c\tau = 90 \text{ mm}$   
Radius 1st layer = 25 mm

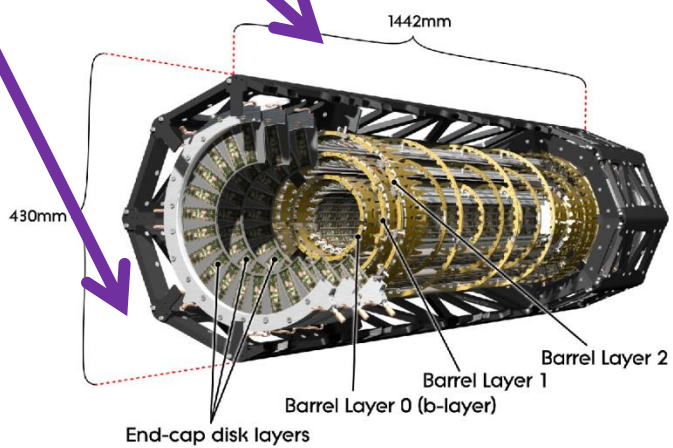
# >2.5 TEV JETS SEEN!







# DEEP IN THE HEART OF ATLAS



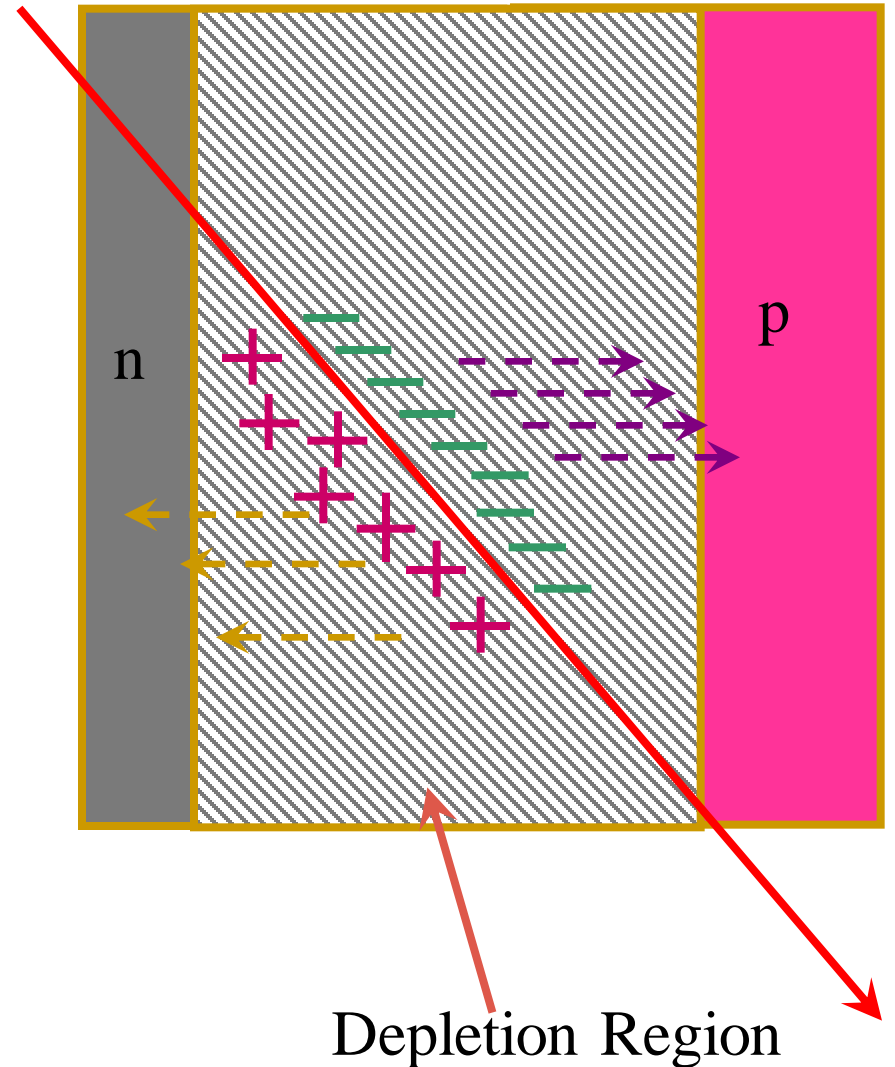
# SILICON VERTEX DETECTOR

## Silicon Detectors

- Reversed biased diodes
- Crank the reverse voltage up until they are fully depleted

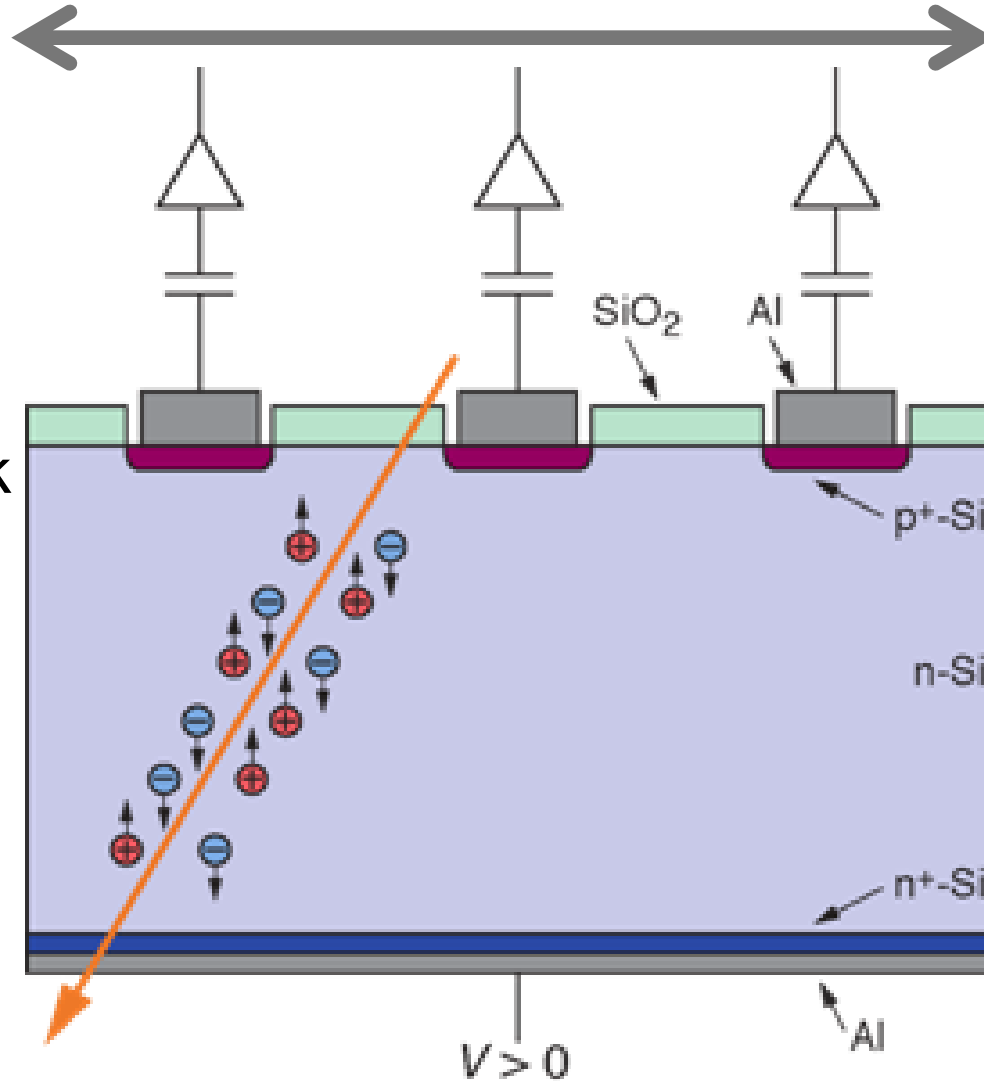
## Charged Particle liberates electron-hole pairs

- Small pulse of current
- *Amplify it and record the hit.*



# PIXEL DETECTOR

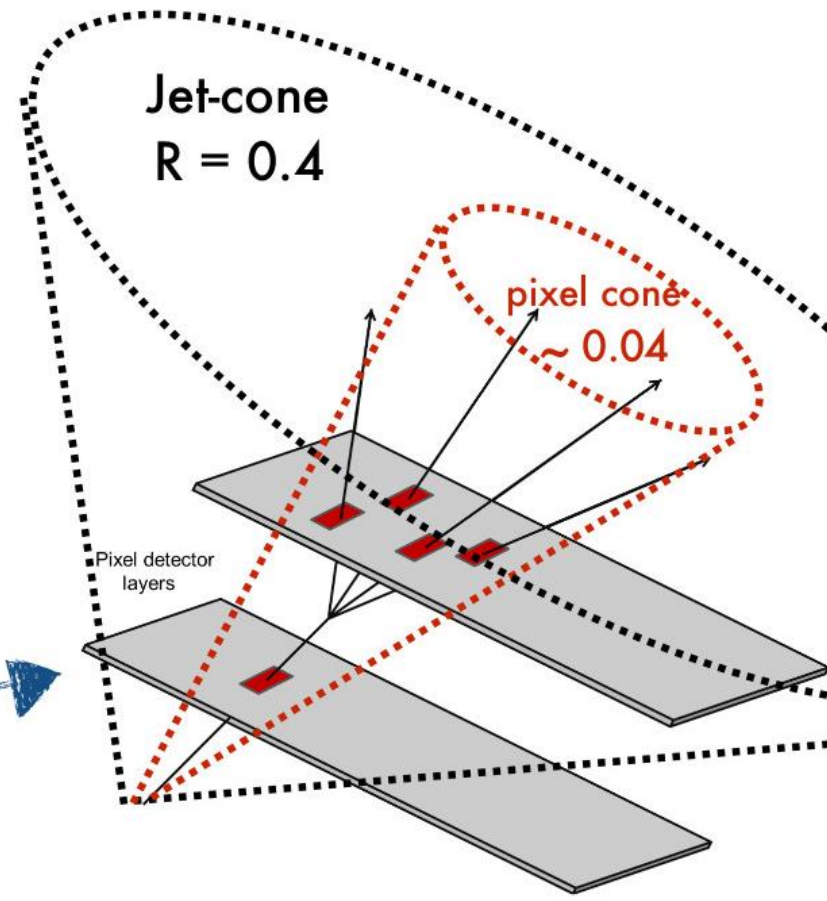
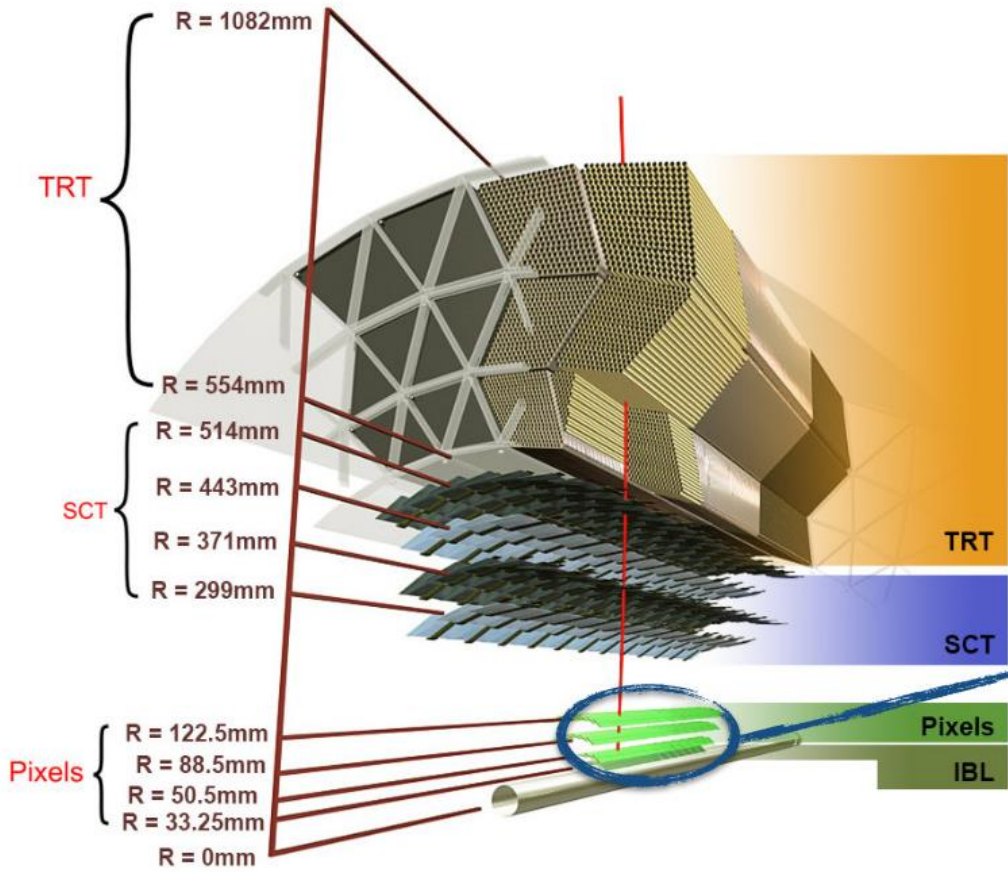
$\sim 150 \mu\text{m}$



Many thousands of tiny sensors

$50 \mu\text{m} \times 250 \mu\text{m}$  and  $300 \mu\text{m}$  thick





### Building the discriminant

$$f_j = \frac{N_{j+1} - N_j}{N_j}$$

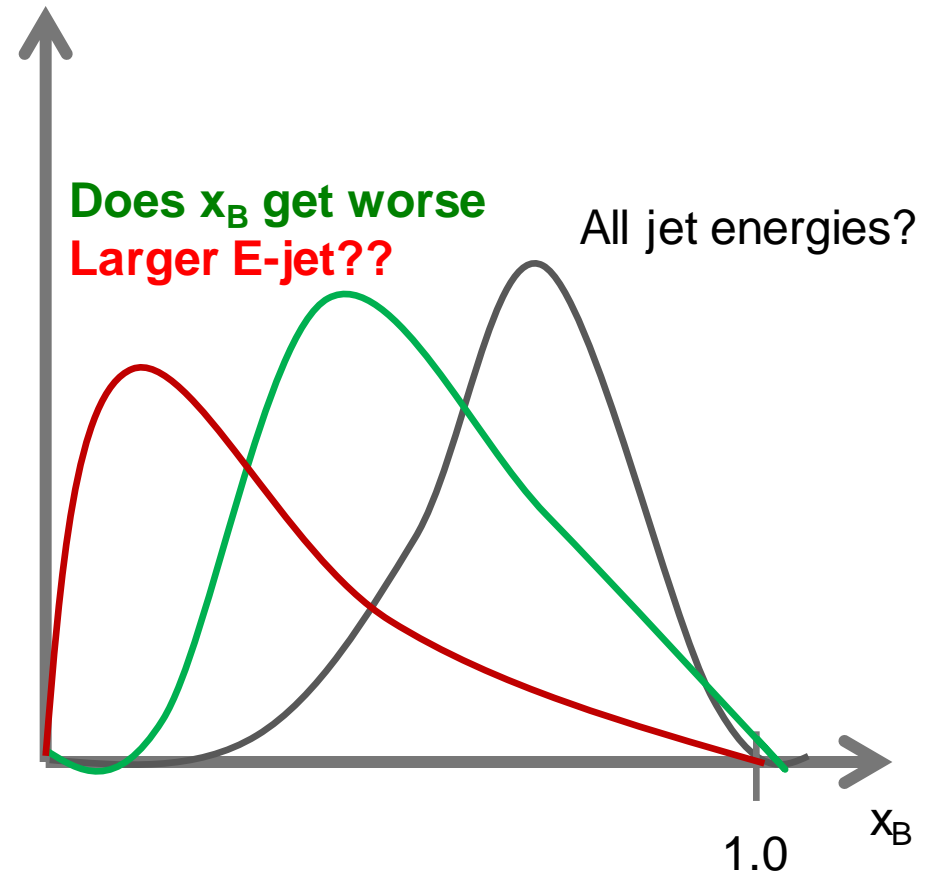
$$Jump = Max(j_1, j_2, j_3)$$

<http://iopscience.iop.org/article/10.1088/0954-3899/43/8/085001/pdf>  
<https://arxiv.org/pdf/1701.06832.pdf>

# ENERGY-FRACTION OF B HADRONS? DOES IT DEPEND ON JET ENERGY?

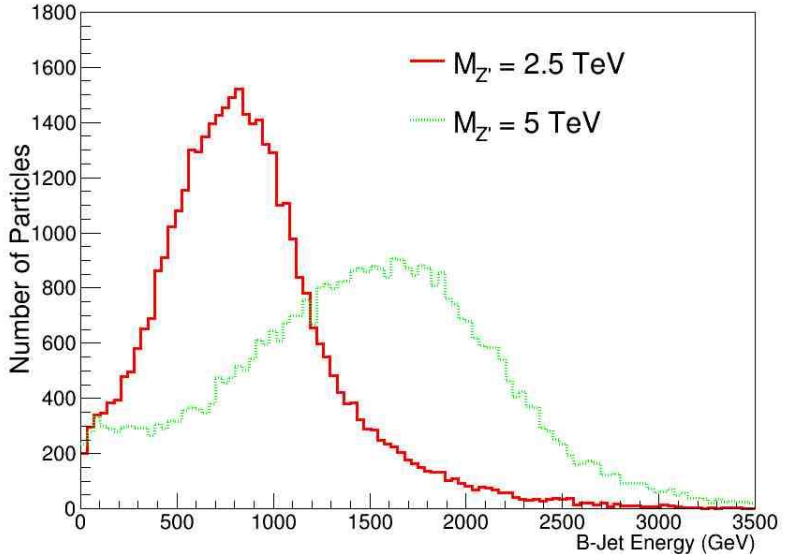
## Energy fraction of B hadrons as Jet energy increases

- $x_B$  Logarithmic?
- $x_B$  Non-linear?
- $x_B$  Constant?
- If Jet energy  $\rightarrow$  more tracks
  - Helps taggers



Let's at least find out what simulations say!

# B HADRON ENERGY FRACTION (PYTHIA+FASTJET)

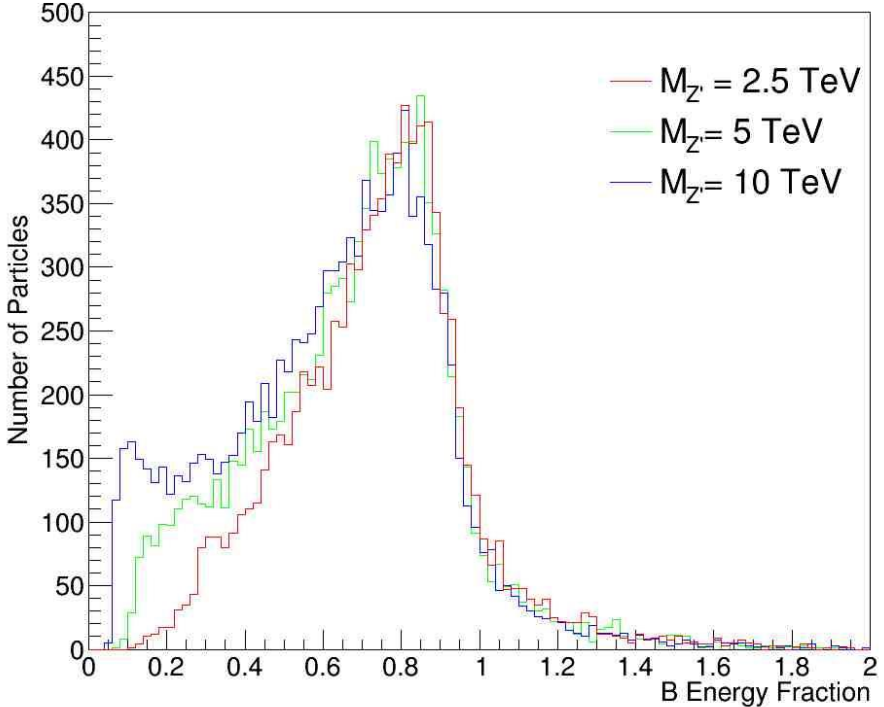


The energy of B hadrons in our simulation

So as Jet energy increases, the problems of B-taggers will get worse.

And more B hadrons will decay inside the detector volume.

However; Fraction of Jet energy each B hadron has does not strongly depend on energy of the jet.



# SIMULATION DETAILS

## Generator level simulation → Pythia 8

- pp collider with  $\sqrt{s} = 13$  TeV
- Generate hard QCD;  $P_T > 700$  GeV
- Use EvtGen to get B hadron decays correct

## Jet simulation → FastJet 3

- Anti- $k_T$  algorithm for forming jets
- Can set jet cone size
  - We've used  $R = 0.2$

## Semi-Toy Detector simulation → GEANT4

- Volume → Cylinder 1.4 m radius filled with air, 2T mag. Field
- Silicon layers
  - Active at radii 25, 50, 88, and 122 mm
    - “Pixels” 50 x 400 x 300  $\mu\text{m}$  ( $\phi$  x z x r)
    - ***inner layer 50 x 250 x 300  $\mu\text{m}$  → IBL-like***
  - *Passive cylinders 2.5 mm thick →  $X_0=2.5\%$  per layer*

# $\Delta$ HIT FRACTION $\rightarrow F_I$

We *define* a quantity we call “Hit-difference Ratio” or “Hit-ratio” for short  $\rightarrow$  “ $f_i$ ”.

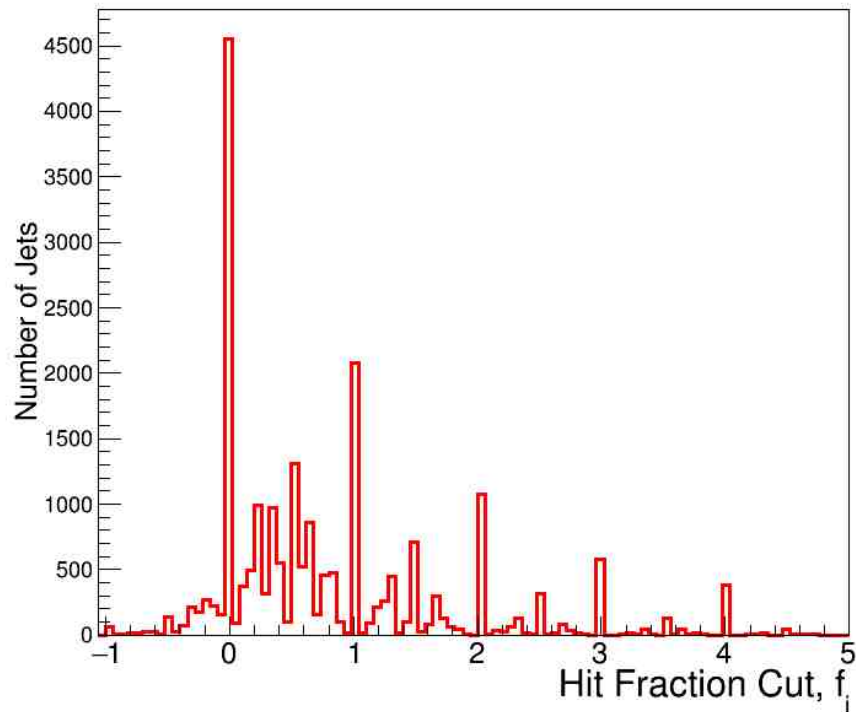
- Use cone  $\Delta R < 0.04$  around jet axis.

**from  $i^{\text{th}}$  layer:**  $f_i = (\#hits_{i+1} - \#hits_i)/\#hits_i$

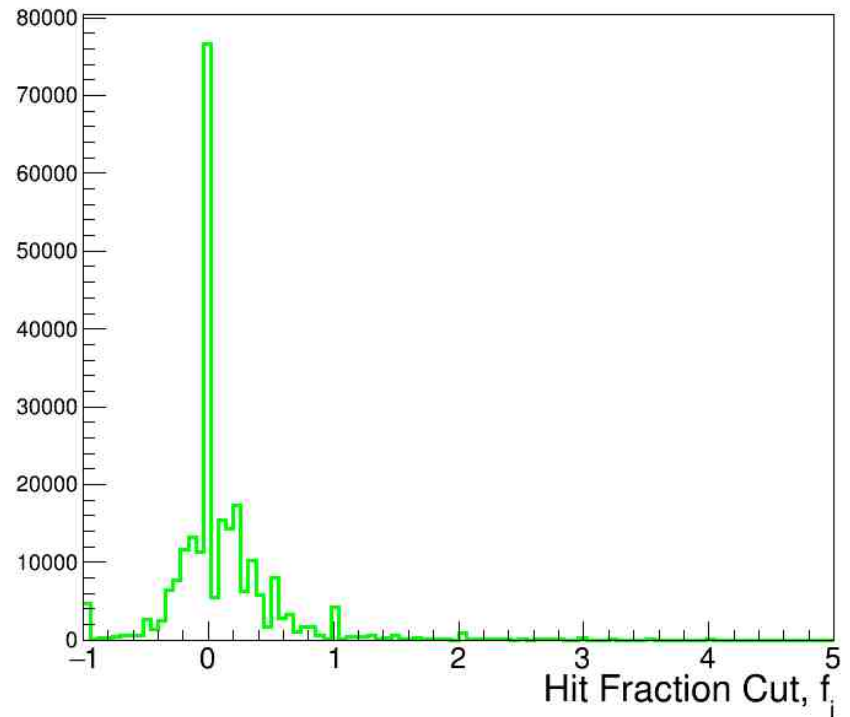
- Can only have positive or zero hits, so:
  - $f_i$  is bounded from below by -1. & unbounded from above.
- 
- Have a look at the  $f_i$  distribution.
    - Note: This sample  $\rightarrow$  0.5 to 2.5 TeV jets.

# $\Delta$ HIT FRACTION = $F_I$

Jets with a B hadron  
All gaps containing a decay



uds Jets

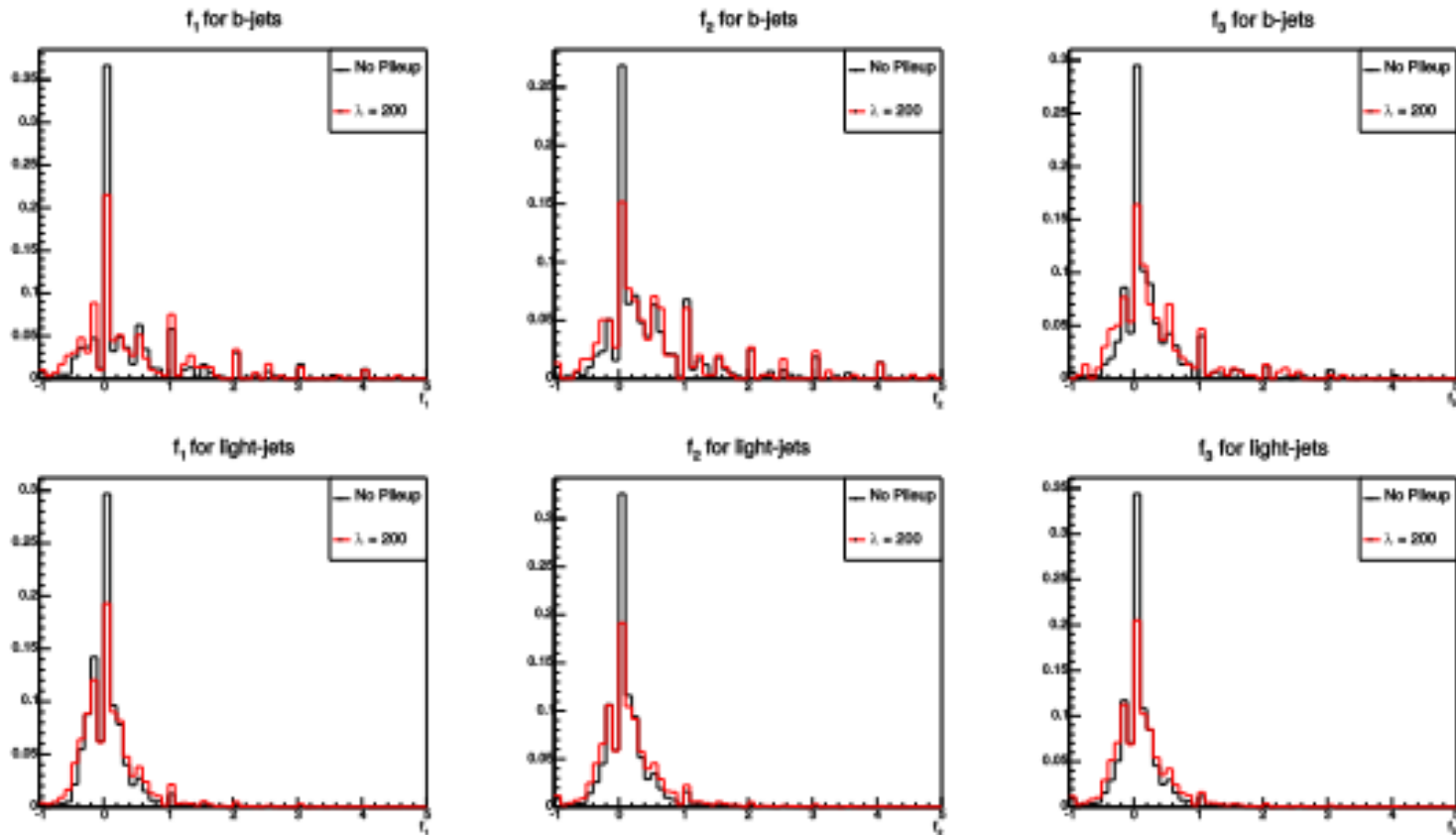


Simulation  $Z' \rightarrow u, d, s, c$ , and b quarks only.

This looked promising  $\rightarrow$  next use it as a cut variable

# Pileup - $f_i$

How does pileup affect our  $f_i$ ?



# Pileup - Conclusion

- Tagger insensitive to increased pileup (tested to pileup with  $\lambda = 200$ ),
  - little change in  $f_i > 0$
- Makes sense - most pileup does not result in hits in  $R < 0.04$  between layers.

**NOTE: All plots now have pileup with  $\lambda = 45$ .**



# APPLYING $\Delta\text{HIT}_F$ CUT

Start at  $f_i = -1.0$  (i.e. no cut at all)

- And Start increasing the cut.

At each cut value, Plot (Number of Events passing cut)/(number of starting events)

**NOTE! Only count B hadron jets where B decayed inside the layers!**

“ALL Layers” plot is logical OR of individual layers

- if any one of the  $\Delta\text{Hit}_f$  between any pair of layers passes the cut, the event passes.

# $F_1$ EFFICIENCY AND PURITY - LAYERS 1234

$M_{Z'} = 2.5 \text{ TEV}$

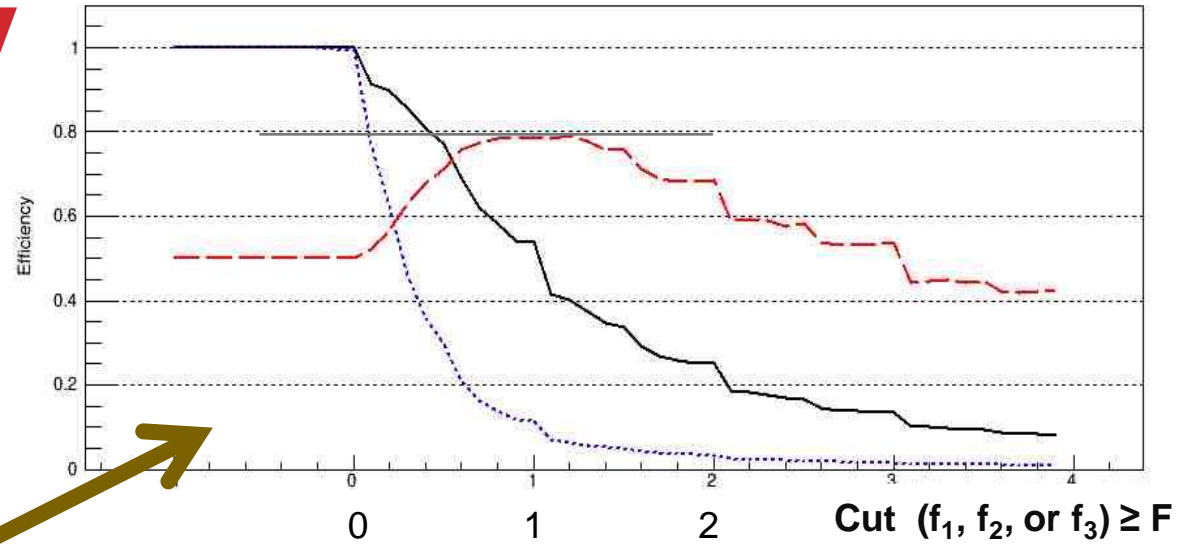
$$f_i \geq F=1$$

Layers:

1-2 or

2-3 or

3-4

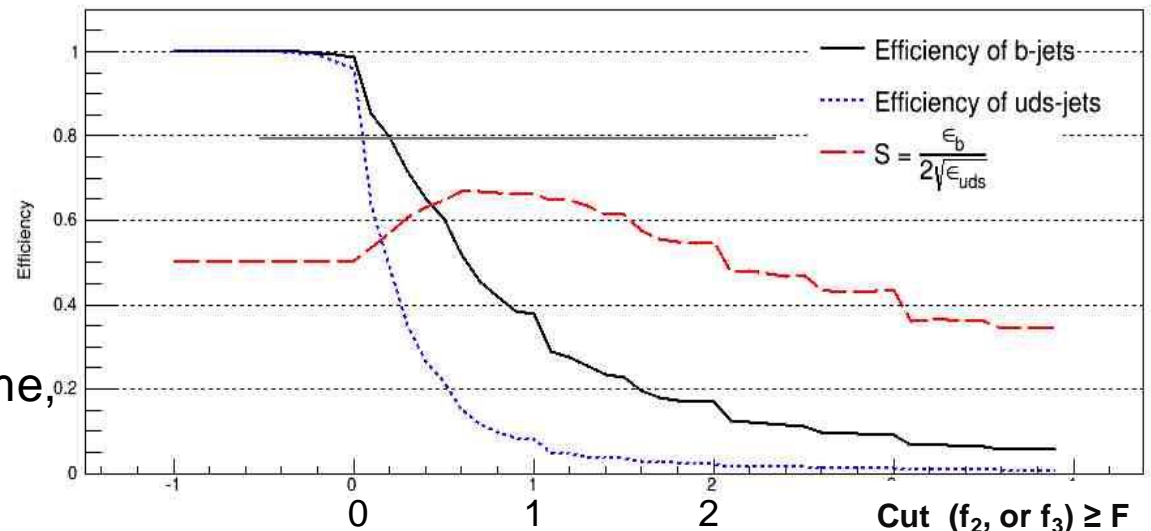


Choose this one!!

Layers:

2-3 or

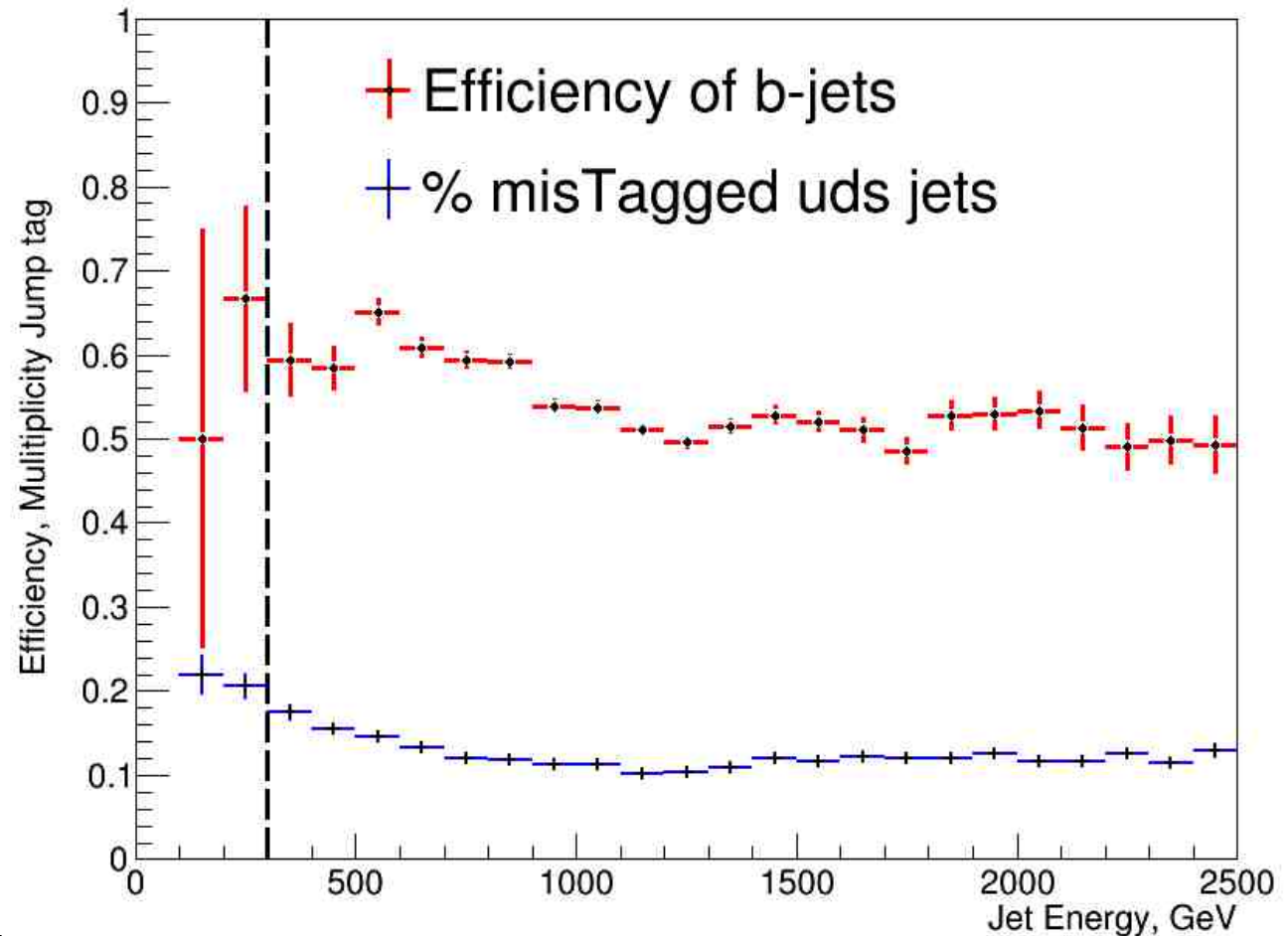
3-4



Fiducial region  $\rightarrow$  Whole volume,  
So can compare.

# EFFICIENCY VS. JET ENERGY

Some uds separation – efficient for high Pt jets



# **ARTIFICIAL NEURAL NETWORK**

# Building the ANN

ROC analysis showed that 8-layer ANN was as effective as a 2-layer ANN → not really a “deep learning” problem, Reducible, small feature set.

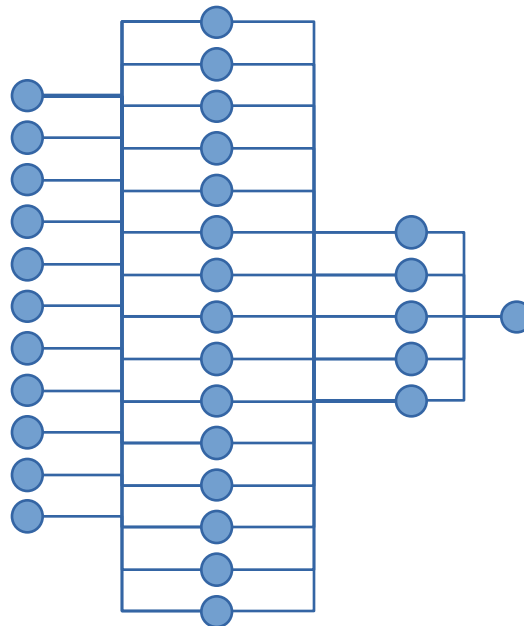
Fewer node layers means less chance of overfitting.

Used **TMVA** for a “standard HEP” packaged ANN.

Activation function was  $\tanh(x)$ .

Input Layer:

- Raw hits in each layer
- Multiplicities  $f_1, f_2, f_3$
- $\max(f_1, f_2, f_3)$
- Jet Energy,  $p_T$  and mass



2 Hidden layers:  
16 and 5 neurons wide.

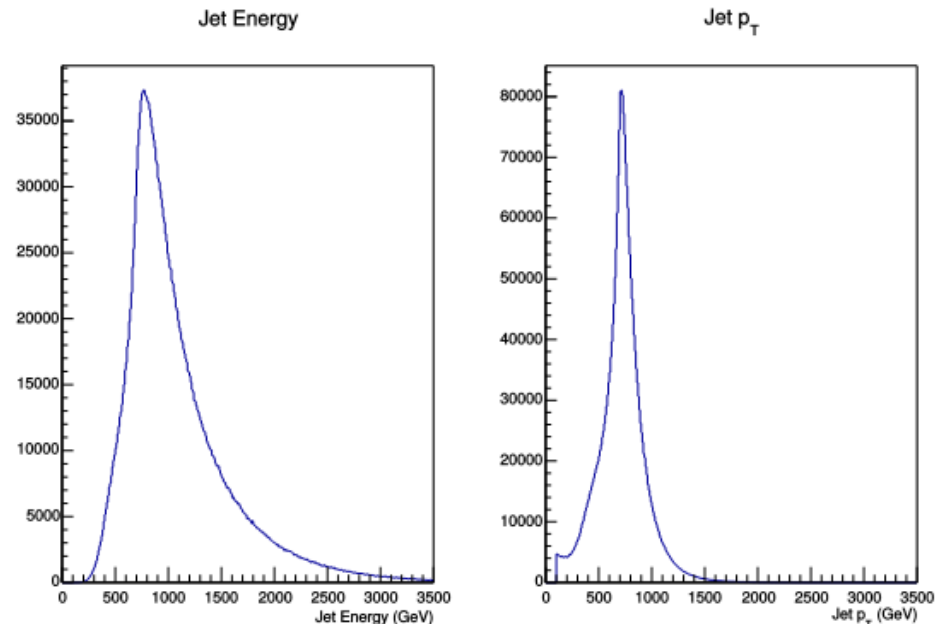
(Smaller second hidden layer helps prevent overfitting).

Architecture → educated guessing.  
Too many combinations to evaluate

# Training Sample

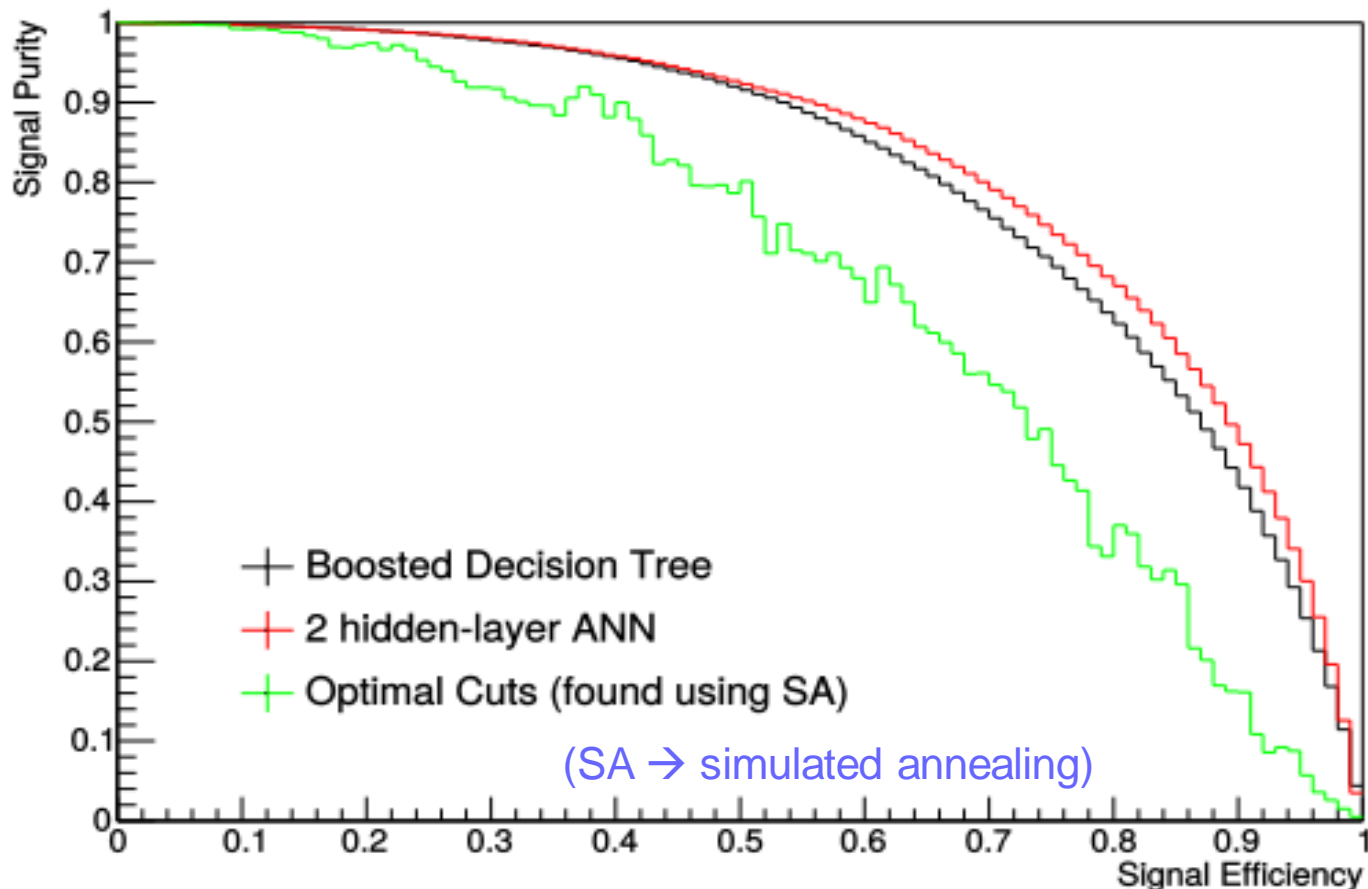
- Training sample had min  $p_T$  of interaction at 700GeV.
- 1 million general hard QCD events (anything goes), 300,000 hard QCD events resulting in B-hadron production.

**Testing is then done on an independent hard QCD sample.**



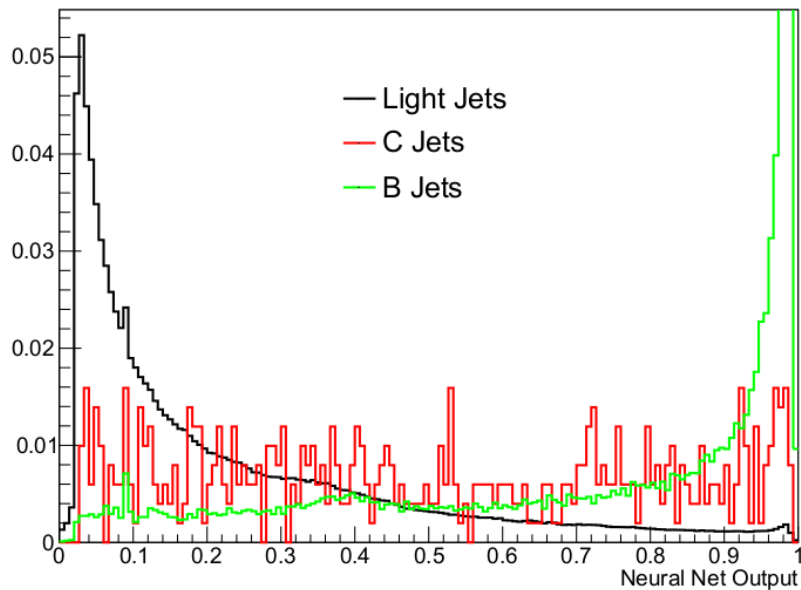
# Choosing a ML technique

- Each ML technique has a distinct set of advantages/disadvantages.
- ROC curve (Signal Purity vs. Signal Efficiency)

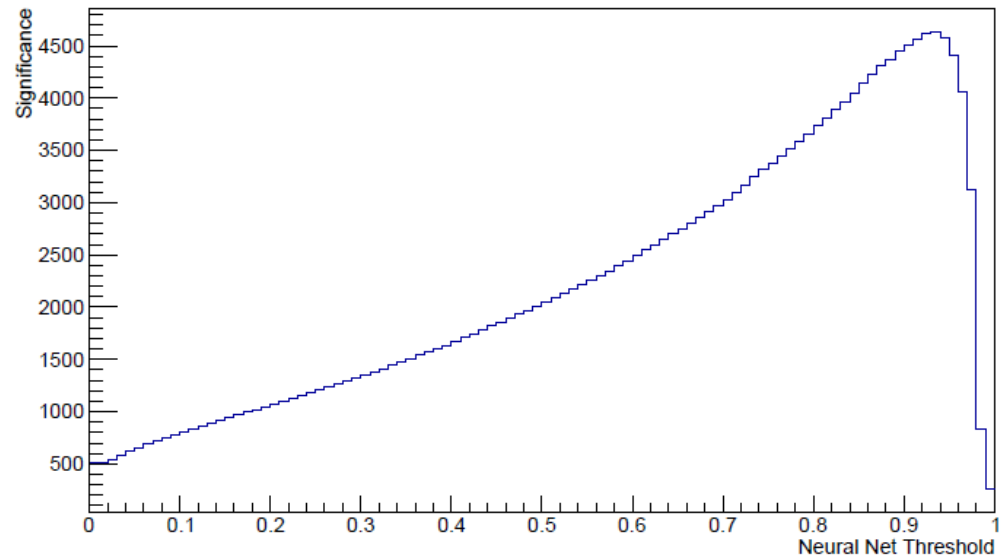


# ANN TEST MC RESULTS

Neural Net Response



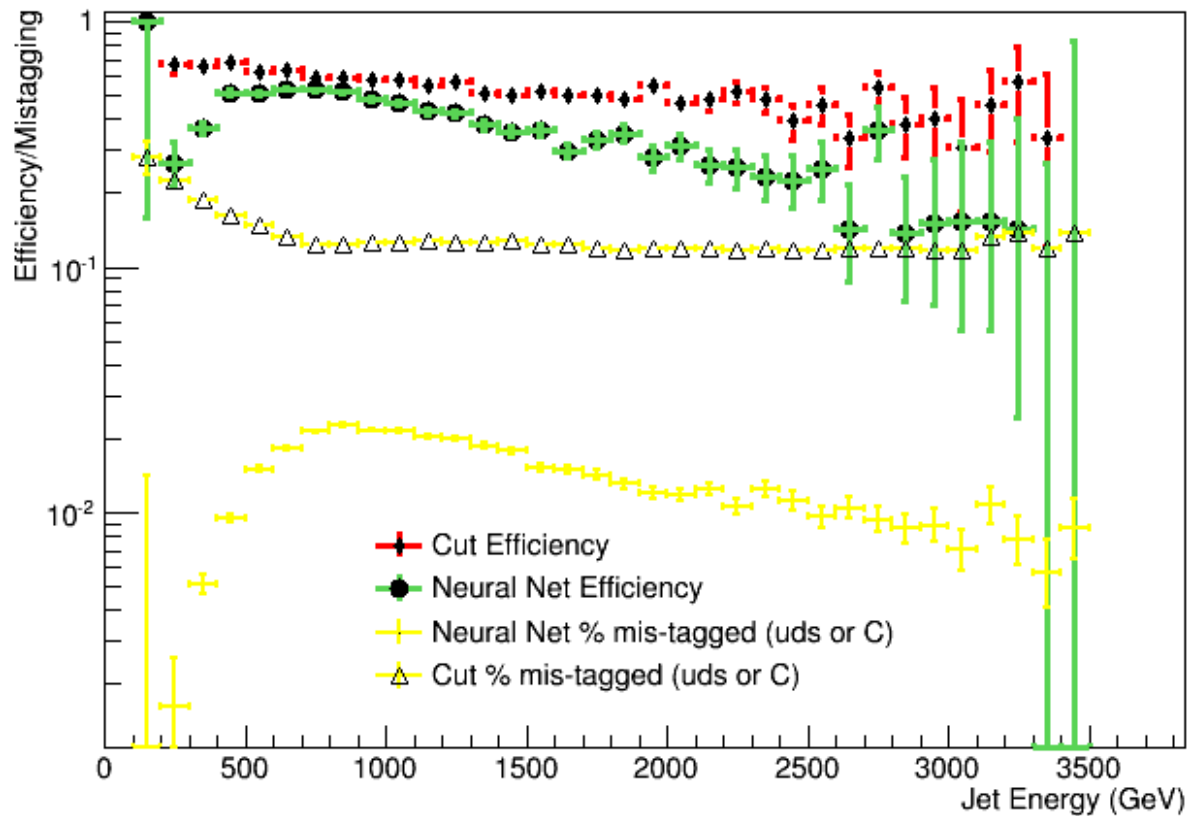
$$\text{Significance} = \epsilon_b / \sqrt{\epsilon_{uds}}$$





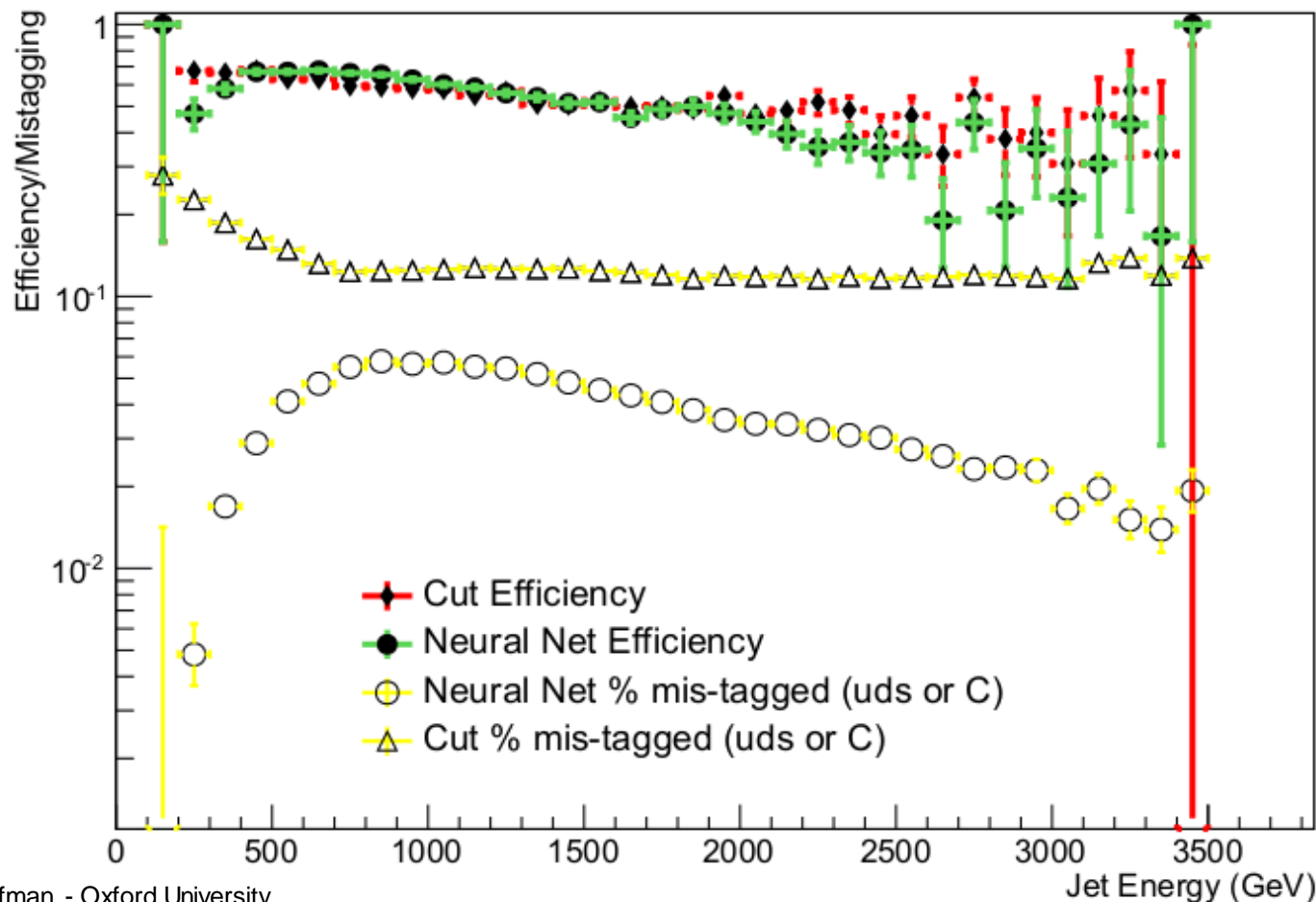
# WITH $\rightarrow$ ANN OUTPUT $> 0.9$

Comparison of Cuts ( $F \geq 1$ ) and ANN



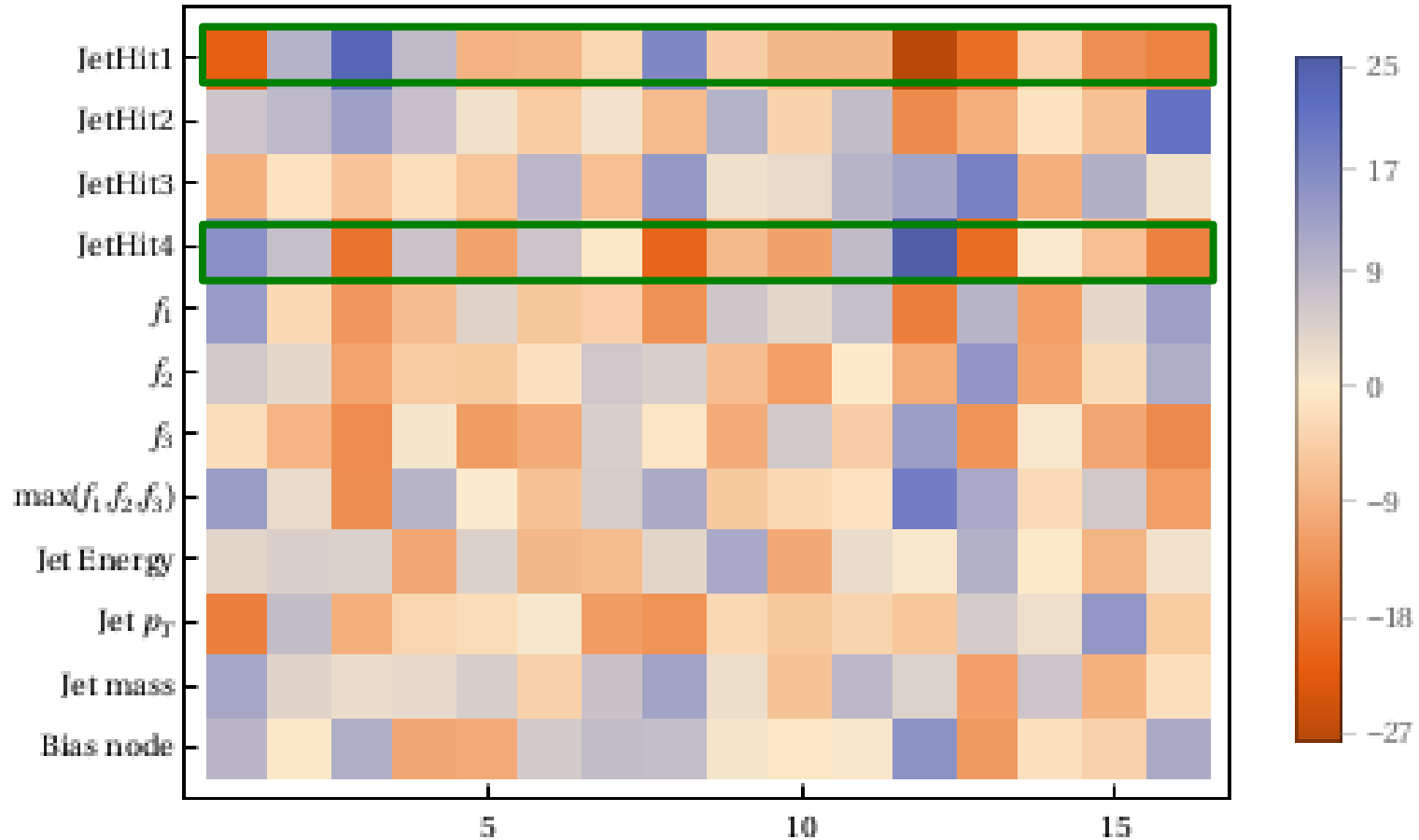
# WITH $\rightarrow$ ANN OUTPUT $> 0.75$

Comparison of Cuts ( $F \geq 1$ ) and ANN



# HEAT MAP

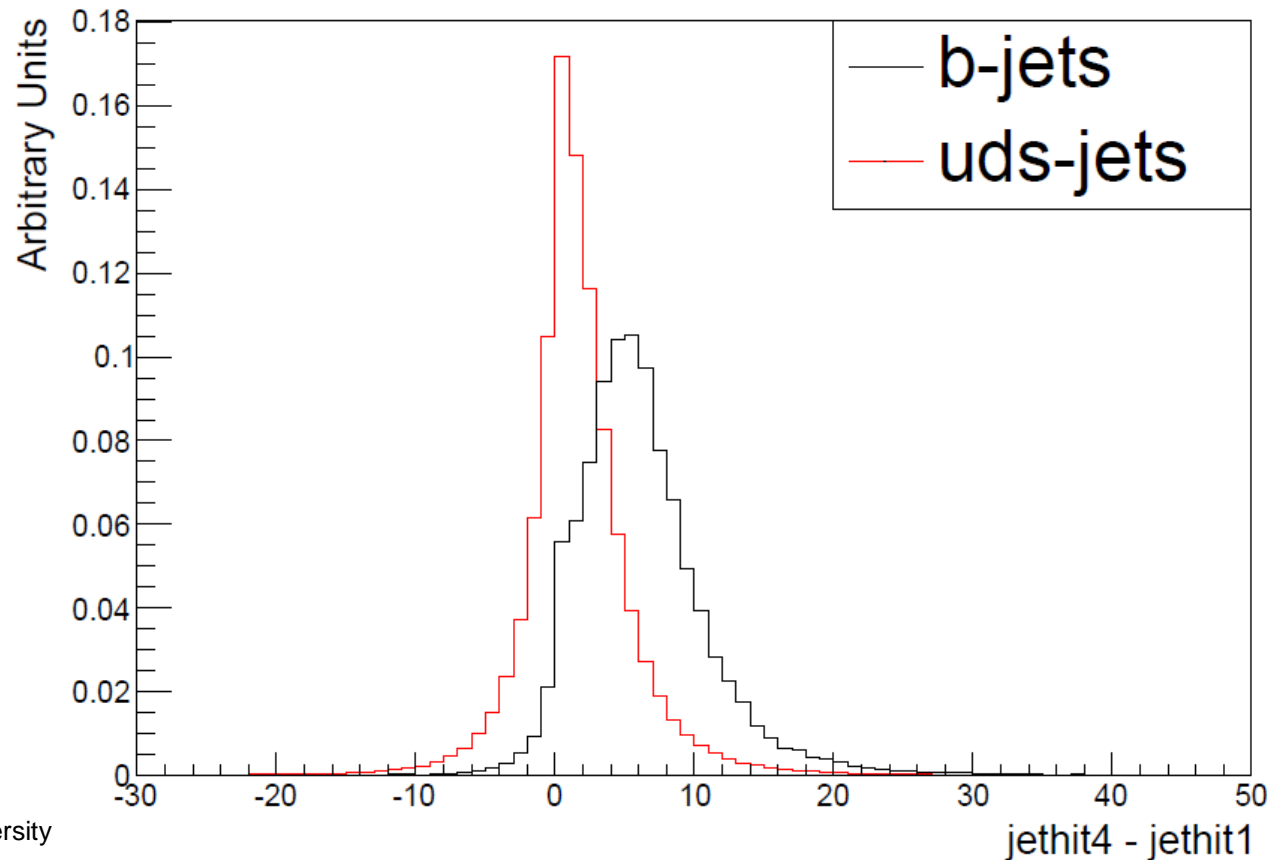
Neural Net seems to find  
 $N_{hit4} - N_{hit1} \rightarrow$  good discriminator



# ANN INSPIRED

Straight hit difference between layer 4 and layer 1.

Definite advantage can be seen.

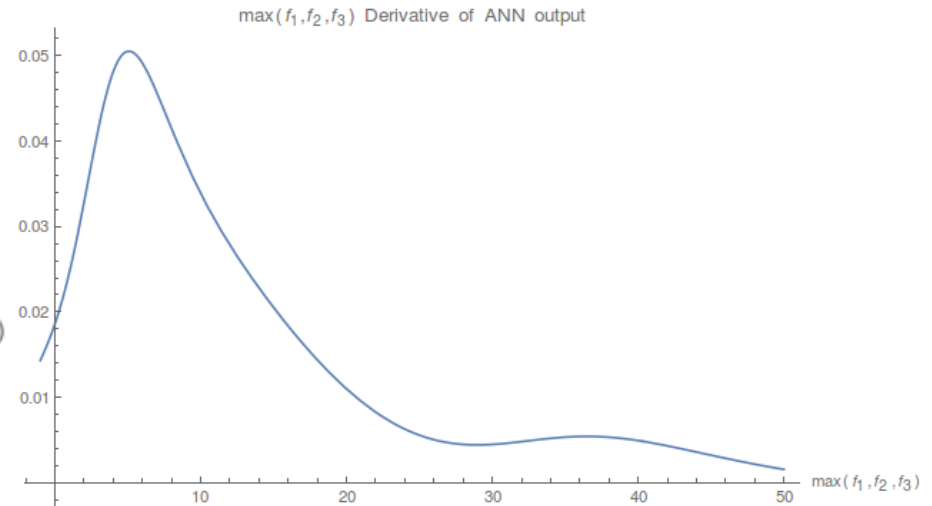
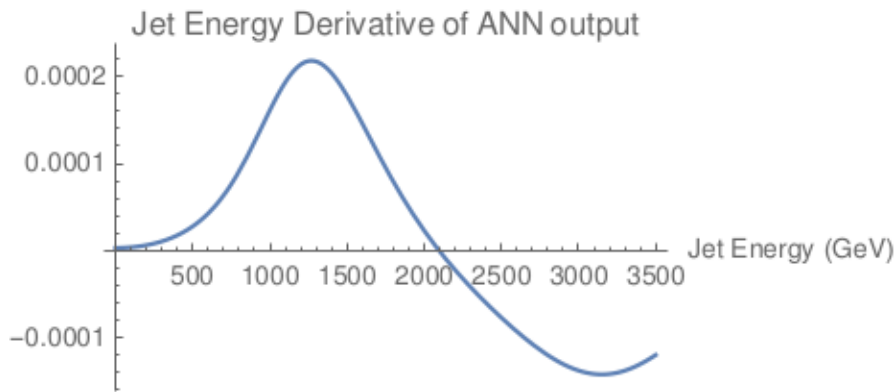


# ANN OUTPUT DEPENDENCE (EXAMPLES)

NN is just a complicated function  $h(x_j)$  – Look at  $\Delta h/\Delta x_j$

## MAX(F)

### JET ENERGY



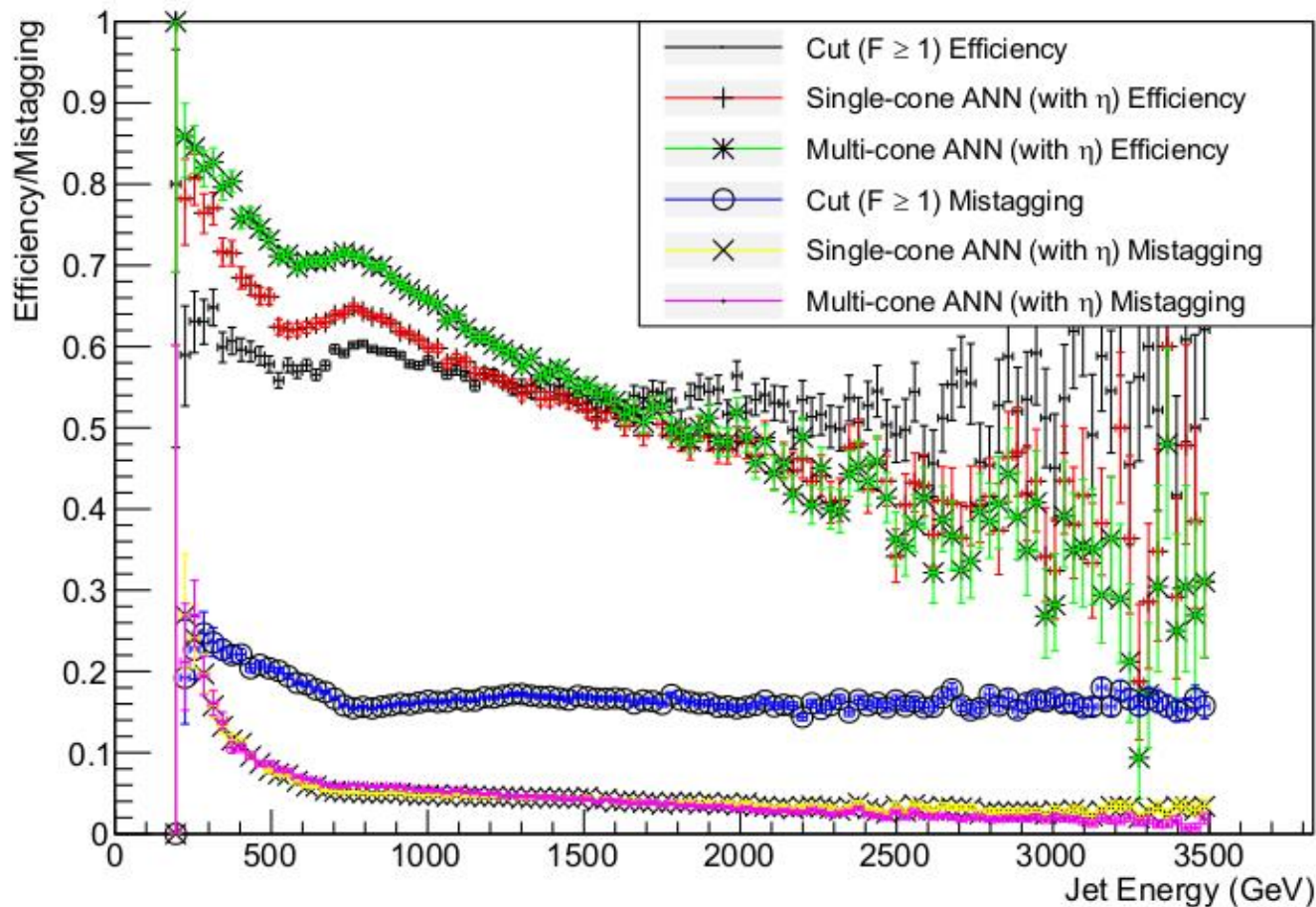
Trained NN output is ~100 times more sensitive to Max(F) than to Jet Energy

**Conclude: The Machine is focused on the “right” types of quantities.**

# LAST THING - MULTIPLE CONE SIZES

Include more B hadron decay products @ lower Energy

$\Delta R < 0.04$ ,  $\Delta R < 0.1$  and  $\Delta R < 0.2$  added  
Each with own set of inputs



Training sample  
Shape dependence?

# CONCLUSIONS

**Using Multivariate techniques significantly improves rejection of uds jets.**

**Results robust to pile-up**

**Multiplicity Jump technique more immune to searchlight effect and highly boosted B hadrons.**

**If continues to show promise, could be added to existing taggers to enhance efficiency at TeV scale jet energies.**

**Experiments (ATLAS, CMS) are looking into the technique with full simulations → Clear next step.**

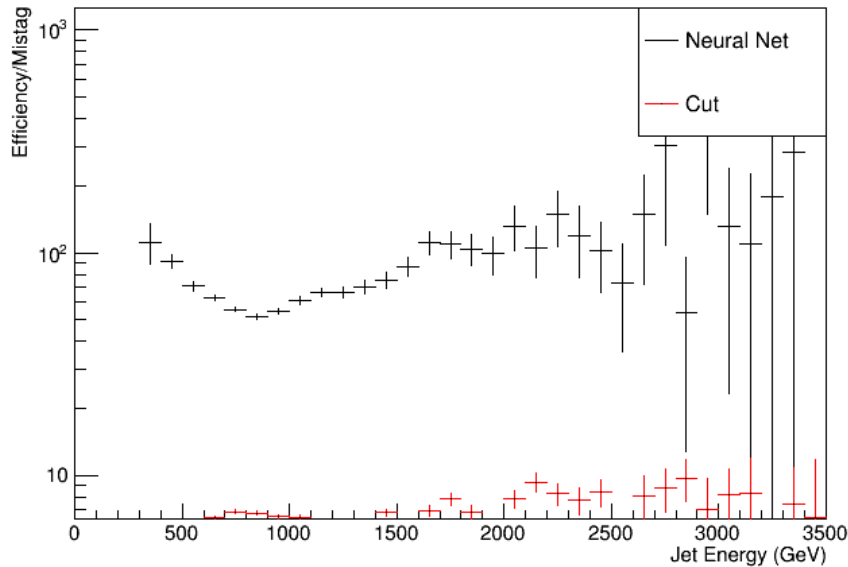
**HAND OFF TO  
ATLAS & CMS!!**

**THANK YOU! (BACKUP SLIDES)**  
B. Todd Huffman - Oxford University

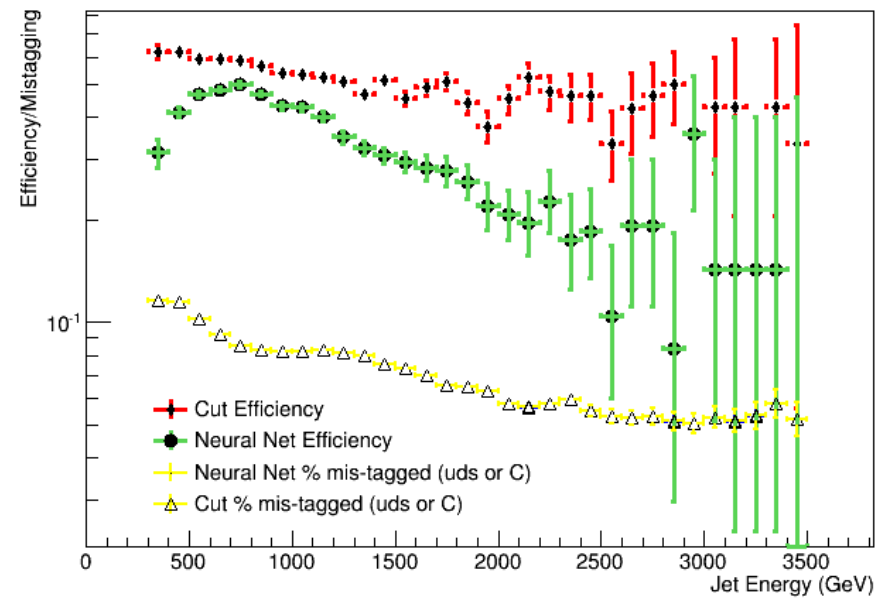


# DEAD MATERIAL REMOVED – NO EARLY SHOWERS

Neural Net Efficiency/Mistag ratio

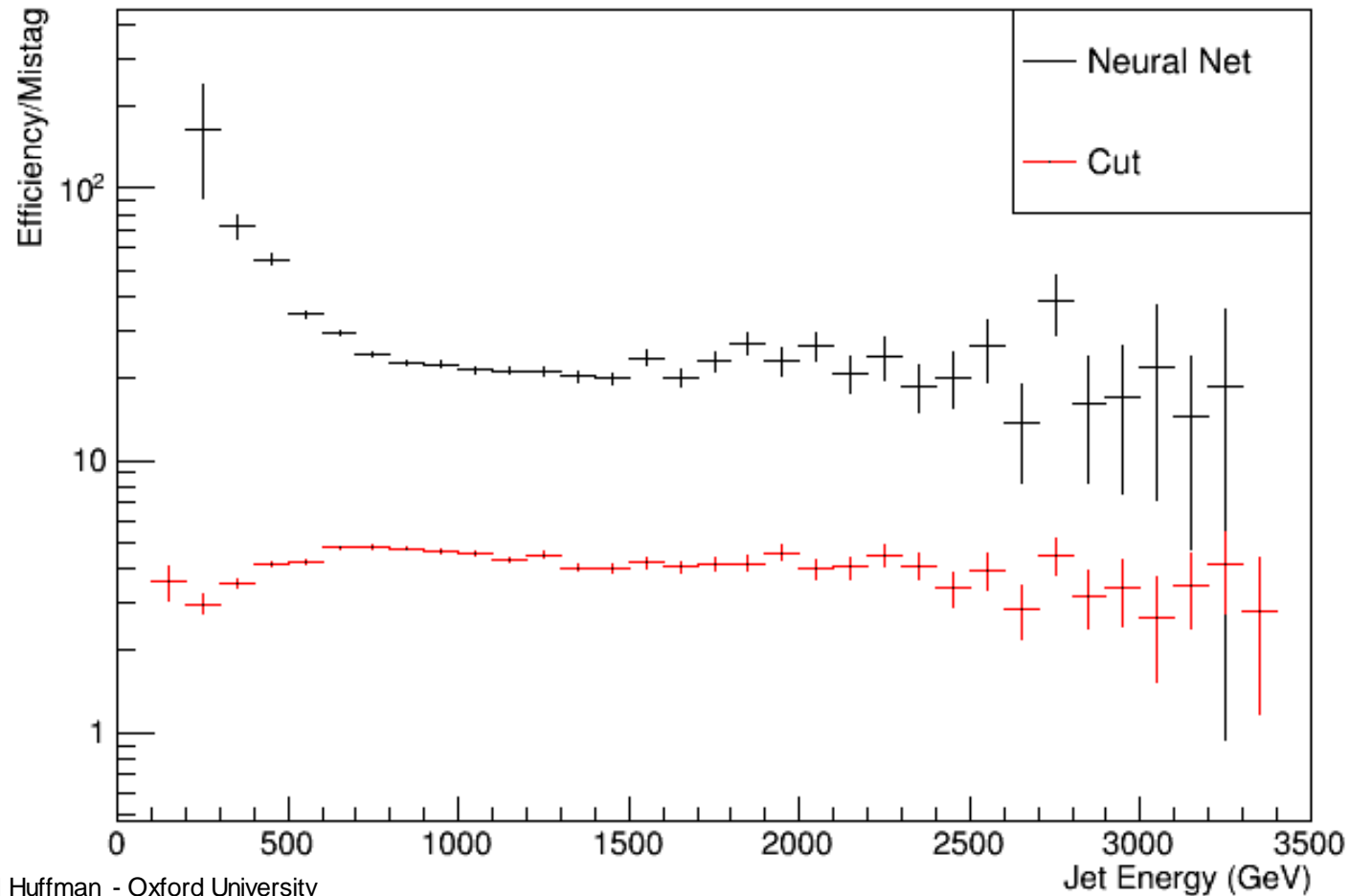


Comparison of Cuts ( $F \geq 1$ ) and ANN



# ANN - STRAIGHT CUT COMPARISON

Neural Net Efficiency/Mistag ratio

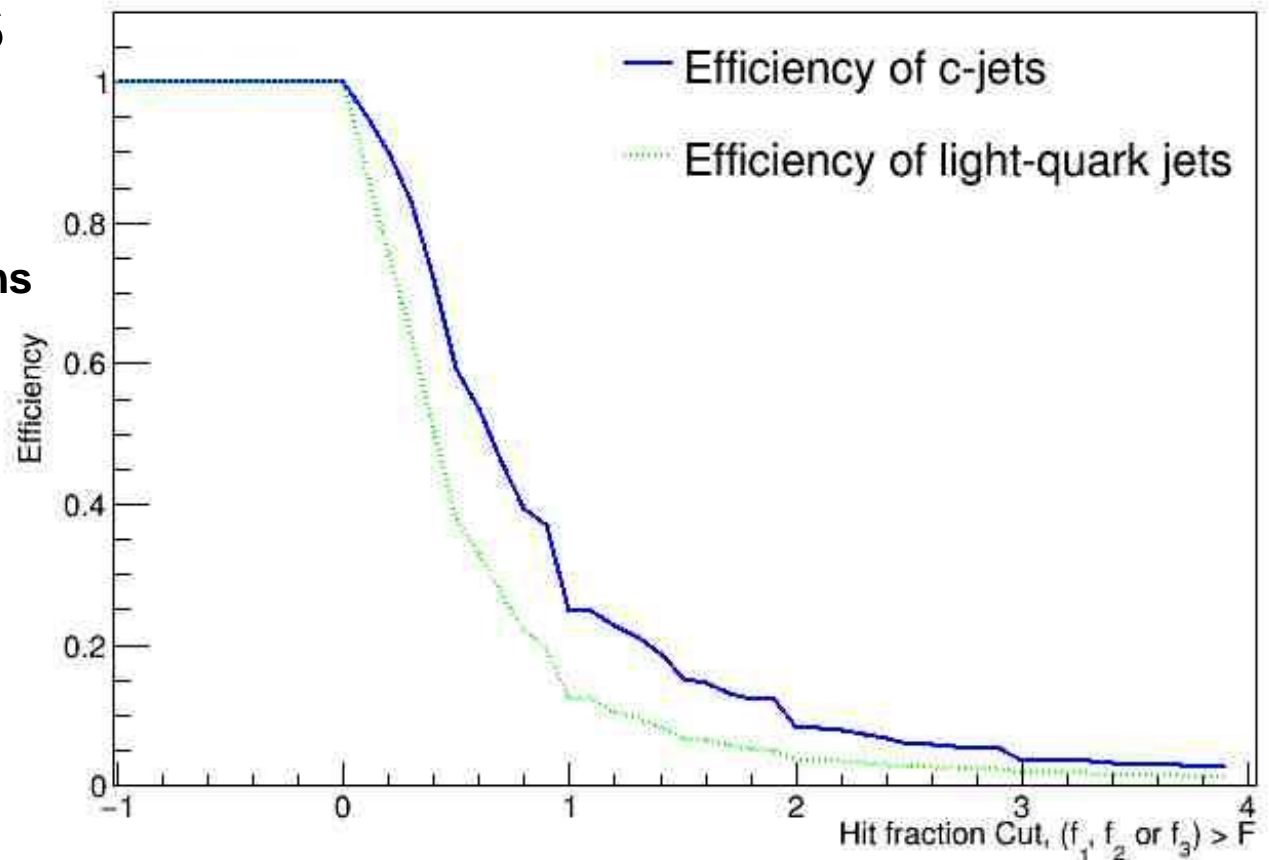


$$\text{MAX}(F_1, F_2, F_3) > \Delta\text{HIT}$$

# Completeness Charm jets

2.5 TeV

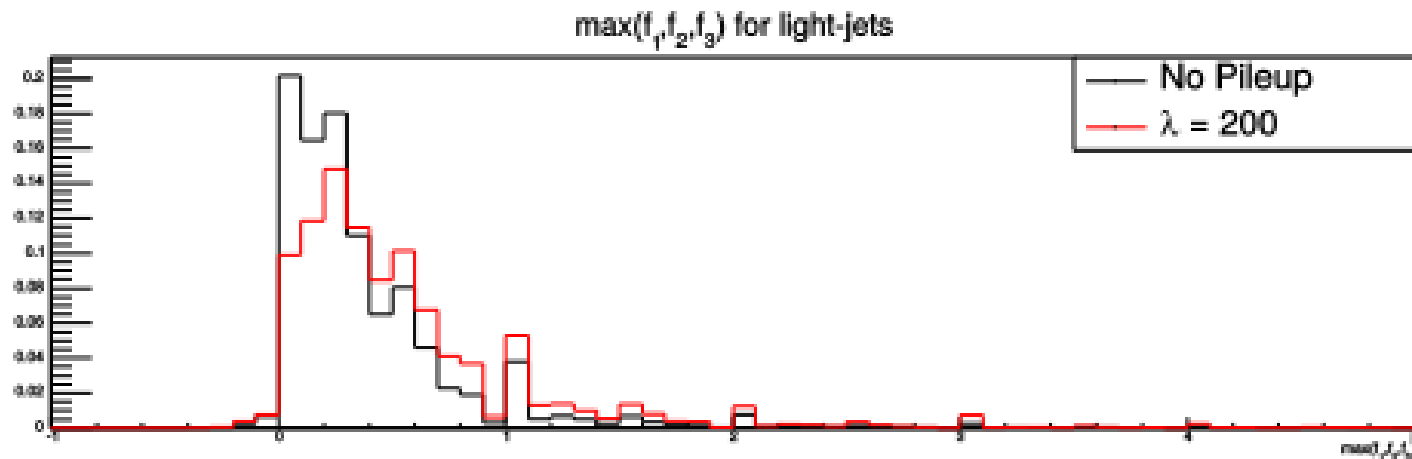
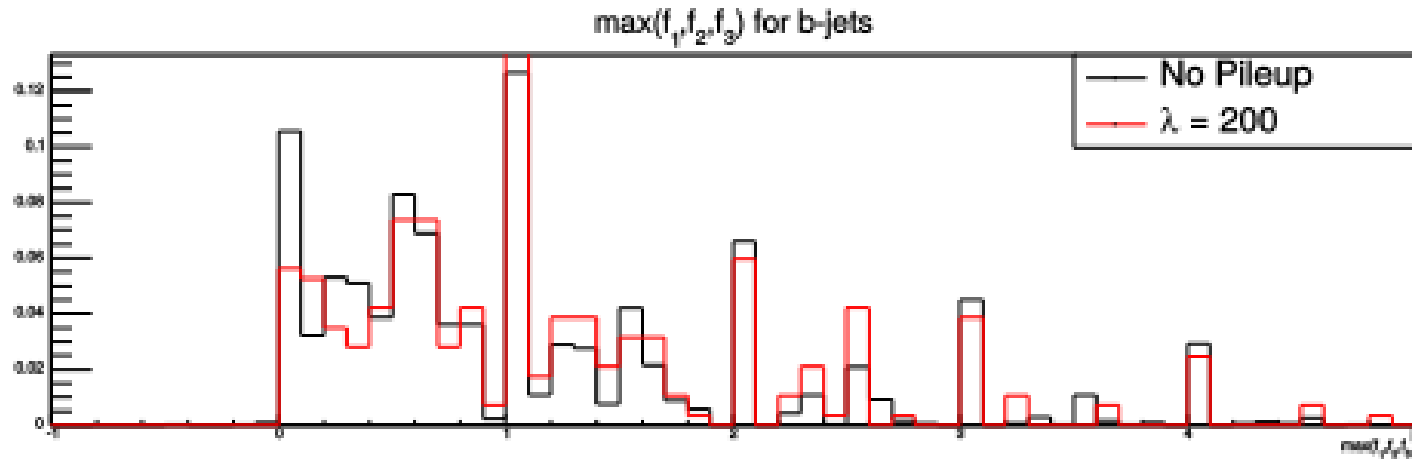
Same set of conditions



# Upgrading the simulation

Making sure it works in a more realistic scenario

# Pileup – $\max(f_1, f_2, f_3)$



# Luminous Region

Events do not always occur at (0,0,0) – modify GEANT4 sim to account for this.

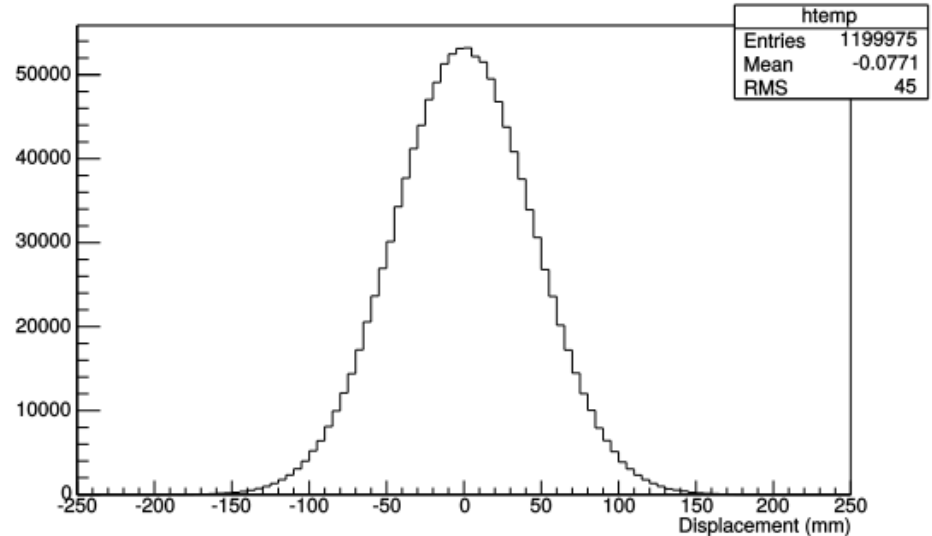
ATLAS has  $\sigma_z = 45\text{mm}$

■

New Problem – Jet vectors no longer from (0,0,0), have to be careful when selecting pixels.

Easily resolved by finding the event primary vertex. Once this is done, luminous region vs origin makes no difference to the quality of the tagger.

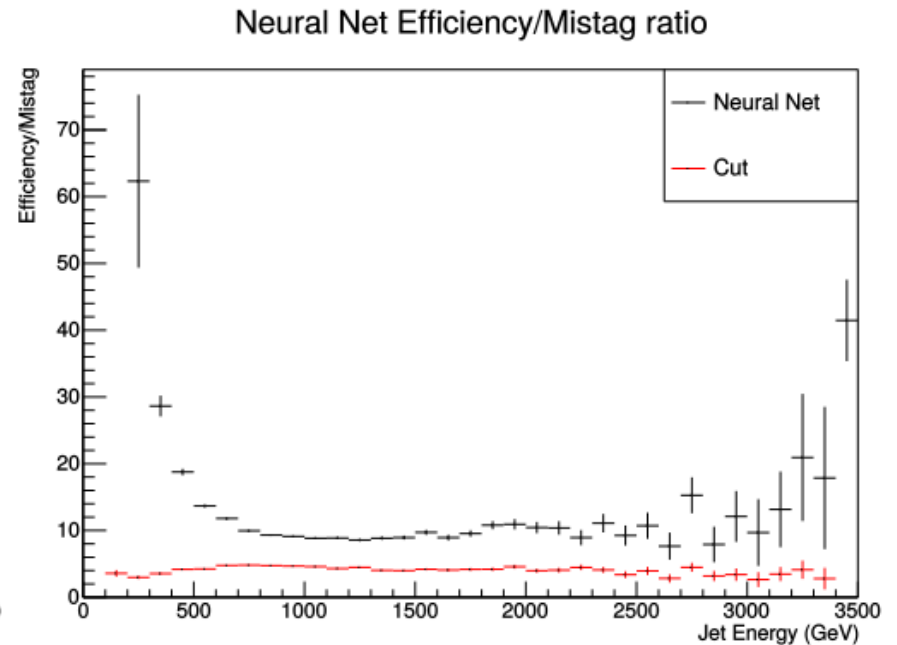
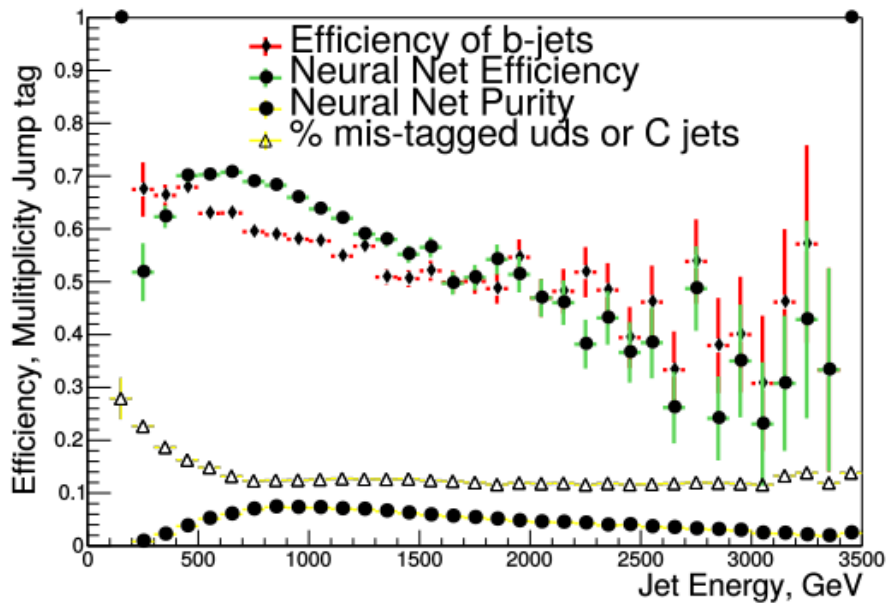
Vertex Z Displacement



# Upgrading the Tagger

We know it works... Let's make it work better!

# How does it do?



Neural Net Wins!!



# Extra Slides

## Energy Deposited in pixels

