

Trigger Level Analysis

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2017 September

A thesis submitted to the University of Manchester
for the degree of Master of Science
in the Faculty of Engineering and Physical Sciences

SUMMARY

Here is my template for PhD or other theses, for pdf \LaTeX (or \LaTeX , but pdf \LaTeX provides better internal hyperlinks).

It is based on the ‘[memoir](#)’ \LaTeX class, which has a lot of useful features/options built-in. The documentation for the memoir class says that ‘[it] provides the functionality of over thirty of the more popular packages, thus simplifying document sources’.

If there is any specific typesetting feature you want to use in your thesis, you should first check in the comprehensive manual for the memoir class via the link above (which has a detailed index). It may well be that what you want is already provided by the memoir class (and it is better to use its built-in capabilities, rather than loading additional style files, unless you have to).

The rest of this template show various examples of features available.

See <http://www.mrao.cam.ac.uk/~dag/THESIS/> for the current version of this template. (This version is V1.12, dated 2016 September).

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DECLARATION

This is the declaration. This is not too long, honest!

ACKNOWLEDGEMENTS

These are the acknowledgements.

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1.1 Standard Model

1.2 Physics of pp Collisions

1.3 Theoretical Predictions

1.4 The Higgs Boson

Detecting the Standard Model Higgs boson is strongly dependent of the predominant production and decay channels for the Higgs boson, which in turn depend on the specifications(?) of the collider used for the search. In this section the relevant production and decay channels at the Large Hadron Collider (LHC) will be discussed.

1.4.1 Higgs Production

While there are many various methods for production of a Higgs boson, at the LHC the cross section is dominated by gloun-gloun fusion (ggF) as shown in figure 1.1, with the second largest cross-section arising from vector boson fusion (VBF). Other significant production processes are the WH/ZH or Higgs-strahlung production modes and associated production with top quarks (ttH) [1].

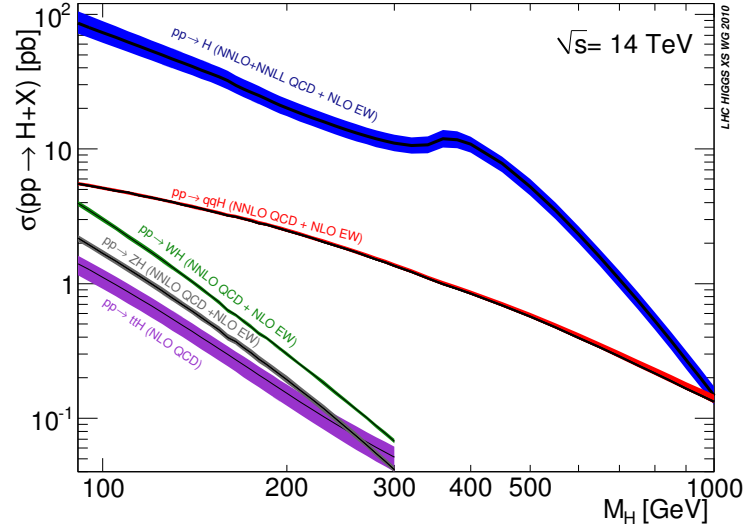


Figure 1.1: SM Higgs Production cross section for $\sqrt{s} = 14$ TeV. $pp \rightarrow H$ corresponds to ggF production and $pp \rightarrow qqH$ VBF. [1]

1.4.1.1 Gluon-gluon fusion

The dominant production mechanism for the Higgs boson in hadron colliders is the $gg \rightarrow H$ production via in intermediate quark loop. The dynamics of this mechanism are controlled by strong interactions, thus calculations of QCD corrections are necessary for any accurate predictions, and have been computed up from next-to-leading order (NLO) to N^3 LO for the ggF process in recent years, along with the inclusion of Electro-Weak corrections in the cross section calculations [1].

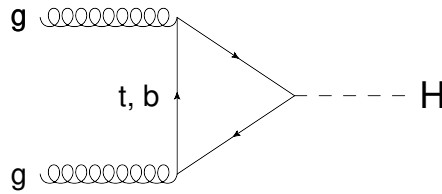


Figure 1.2: Lowest order Feynmann diagram contributing to $gg \rightarrow H$.

1.4.1.2 Vector Boson Fusion

Production of a Higgs boson from the fusion of vector bosons radiated from initial-state quarks is the second largest cross-section at the LHC, as is useful as a production mode due to topological characteristics which can distinguish the event from ggF. In VBF, the Higgs boson is produced along with two jets in the forward regions of the detector, which originate from the initial

Maybe need to link
to the detector
section

quarks as shown in Figure 1.3. In addition central jet activity is suppressed due to the lack of colour exchange between quarks [2]. These distinct features mean that while the cross section for VBF at a Higgs mass of < 200 GeV is dominated by ggF, the easy to detect signature means the channel is a cornerstone of searches for the Higgs boson.

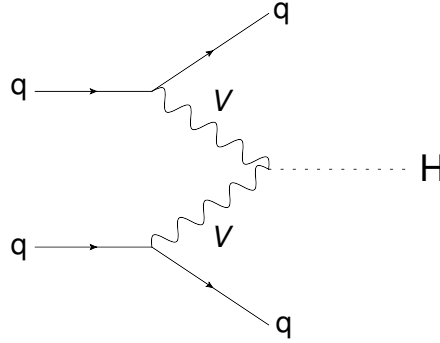


Figure 1.3: Feynmann diagram for the production of a Higgs boson via vector boson (V) fusion, where q denotes any quark or antiquark

1.4.2 Higgs Decay

The branching ratios for decays of the Higgs boson in the Standard Model have been extensively determined using Monte-Carlo event generators. As is to be expected, the relative cross-sections of the decay modes are strongly dependent on the mass of the Higgs boson, as highlighted in Figure 1.4.

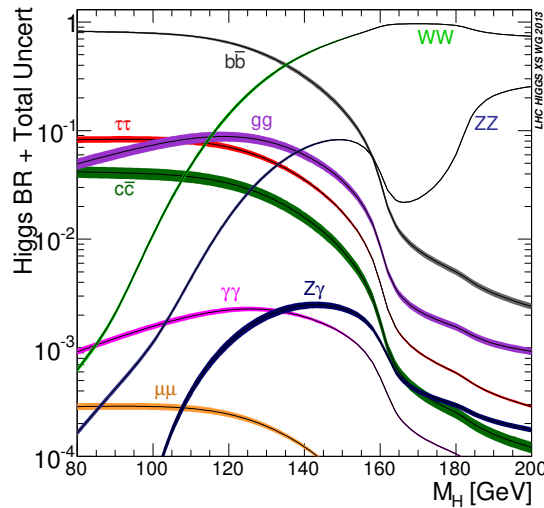


Figure 1.4: Higgs decay branching ratios for the low mass region with their uncertainties [?].

While observations consistent with the Standard Model Higgs boson have been made for the $H \rightarrow \gamma\gamma$, $H \rightarrow ZZ$, $H \rightarrow W^+W^-$ and $H \rightarrow \tau^+\tau^-$ channels, observation of the $H \rightarrow b\bar{b}$ decay channel is significantly hindered owing to the large background from multijet production (Section ?? maybe?) in hadron collisions. Despite this, the topology of the VBF production mechanism makes it a viable option for observation of the $b\bar{b}$ decay channel.

1.4.3 Vector Boson Fusion

CHAPTER 2

DETECTOR

2.1 The Large Hadron Collider

2.2 The ATLAS Detector

2.3 Triggers

2.4 Object Reconstruction

2.4.1 Jets

2.4.2 b -jets

2.5 b -Tagging

Identification of b -quark jets in ATLAS is based on combining the output of three separate b -tagging algorithms: Impact Parameter based (IP2D and IP3D, described in Section 2.5.0.1), Secondary Vertex based (SV, described in Section 2.5.0.2) and Decay Chain based (JetFitter, described in Section 2.5.0.3) into a multivariate discriminant (MV2, covered in Section 2.5.1) which is used to distinguish the jet flavours. These algorithms have undergone continuous improvement over the Run-2 cycle of the LHC to improve the separation of jet flavours.

2.5.0.1 IP2D and IP3D: Impact Parameter based Algorithms

The typical topology for a b -hadron of a secondary vertex displaced from the hard scatter interaction point as a result of the lifetime of b -quark is used as the basis of these algorithms. Impact parameters of tracks from the secondary vertex are computed with respect to the primary vertex of the interaction. The IP2D algorithm uses a transverse impact parameter (d_0) defined as the distance of closest approach of a track to the primary vertex in r - ϕ plane around the vertex. The IP3D algorithm uses both the transverse and a correlated longitudinal impact parameter ($z_0 \sin \theta$), defined as the distance between the point of closest approach in r - ϕ and the primary vertex in the longitudinal plane. . These parameters typically have large values as a result of the lifetime of b -quark . The signs of the impact parameters are also defined to take account of if they lie in front or behind the primary vertex with respect to the jet direction, with secondary vertices occurring behind the primary vertex normally due to background.

To do: I kind of want a diagram here, but that doesn't appear to be the norm

The significance of the impact parameter values ($\frac{d_0}{\sigma_{d_0}}, \frac{z_0}{\sigma_{z_0 \sin \theta}}$) for each track are compared to probability density functions obtained from reference histograms derived from Monte Carlo simulation, with each track being compared to a selection of reference track categories. This results in weights which are combined using a log-likelihood ratio (LLR) discriminant to compute an overall jet weight separating the b , c , and light-jet flavours from each other. [3, 4]

2.5.0.2 SV1: Secondary Vertex Finding algorithm

The secondary vertex algorithm uses the decay products of the b -hadron to reconstruct a distinct secondary vertex. The algorithm uses all tracks that are significantly displaced from the primary vertex associated with the jet, forming vertex candidates for all pairs of track, while rejecting any vertices that would be associated with decay of long lived particles (e.g. K_s , Λ), photon conversions or interactions with the material in the detector. The tracks forming these vertex candidates are then iteratively combined and refined to remove outliers beyond a χ^2 threshold leaving a single inclusive vertex.

The properties of this secondary vertex are used to differentiate the flavour of the jet. The SV1 algorithm is based on a LLR formalism similar to the IP algorithms, and makes use of the invariant mass of all charged tracks used to reconstruct the vertex, the number of two track vertices and the ratio of the invariant mass of the charged tracks to the invariant mass off all tracks. In addition the algorithm is signed in a similar fashion to the IP algorithms and uses the ΔR between the jet direction and secondary vertex displacement direction in the LLR calculation. The algorithm uses distributions of these variables to distinguish between the jet flavours. [3, 4]

Might be worth mentioning the way these are trained

2.5.0.3 JetFitter: Decay Chain Multi based Algorithm

The JetFitter algorithm exploits the topological structure of weak b -hadron and c -hadron decays inside the jet to reconstruct a full b -hadron decay chain. A Kalman filter is used to find a common line between the on which lie the b , c and primary vertices to approximate the b -hadron flight path. A selection of variables relating to the primary vertex and the properties of the tracks associated with the jet are used as input nodes in a neural network. This neural network uses the input variables and p_T and $|\eta|$ variables from the jets, reweighted in the kinematic variables to ensure the spectra of the kinematics are not used in the training of the neural net. The neural network outputs a discriminating variable relating to each jet flavour which are used to tag the jets. [?]

To do: Either understand or just cite

2.5.1 Multivariate Algorithm

The output variables of the three basic algorithms described prior are combined as input into the Multivariate Algorithm MV2. MV2 is a Boosted Decision Tree (BDT) algorithm which has been trained on $t\bar{t}$ events to discriminate b -jets from light and c -jets. The algorithm makes use of the jet kinematics in addition to the tagger input variables to prevent the kinematic spectra of the training sample from being used as discriminating factor. The MV2 algorithm is an revised version of the MV1 algorithm used during Run-1 of the LHC, and has three sub-variants (MV2c00, MV2c10, and MV2c20) of the algorithm distinguished by the exact background composition of the training sample. The naming convention initially referred to the c -jet composition of the training sample, e.g. for MV2c20 the b -jets are designated as signal jets where a mixture of 80% light jets and 2% c -jets was designated as background.

Why?

list all input variables?

The MV2 algorithm has a set of working points, defined by a single value of the output distribution of the algorithm, which are configured to provide a specific b -jet selection efficiency on the training $t\bar{t}$ sample. Rather than being used independently, physics analyses will make use of several working points as an increase in b -jet efficiency (corresponding to *looser* b -jet selection) will bring an increased mistag rate of light and c -jets.

These algorithms were refined prior to the 2016 Run-2 data-taking session in response to c -jets limiting physics analyses more the light-jets. This change to enhance the c -jet rejection meant that for the MV2c10, the c -jet fraction was set to 7% in training and the fraction for MV2c20 was 15%. There were a selection of other improvements to the algorithm made to the algorithm relating to the BDT training parameters and the use of the basic algorithms before the 2016 data taking. With these refinements, the MV2c10 algorithm was found to provide a comparable level of light-jet rejection to the original 2015 Mv2c20 algorithm with improved c -jet rejection, so was chosen as the standard algorithm for 2016 analyses. [3]

EVENT SELECTION

This section describes the selection criteria required for the events and reconstructed objects used in the analysis. These cuts and criteria are designed with the VBF $H \rightarrow b\bar{b}$ event topology in mind, along with the limitations introduced by considering the available trigger chains as discussed in Section ???. These cuts are applied in the VBF $H \rightarrow b\bar{b}$ analysis and the direct object comparison covered in Chapter 4.

3.1 Events

Data events were required to pass the all year 25ns Good Runs List^a and also be Clean ??.

3.2 Offline Jets

Offline jet reconstruction was performed by the anti- k_t algorithm ($R=0.4$) as discussed in Section ??. Jets were calibrated in line with the 20.7 recommendations ??. When considering individual jets during the analysis, all jets were required to have a $p_T > 45$ GeV to be recorded.

3.3 Online Jets

Online Jet reconstruction is a mystery. A full collection of online jets was recovered by extracting the split jets (Section

^adata16_13TeV.periodAllYear_DetStatus-v88-pro20-21_DQDefects-00-02-04_PHYS_StandardGRL_All_Good_25ns.xml

3.4 Offline b -jets

The specifics of b -tagging are covered in Section 2.5. Offline b -jets were tagged using the $MV2c10$ -tagger^b with two defined efficiency working points: *Tight*, with an overall efficiency of 70% and *Loose* with 85% tagging efficiency.

3.5 Online b -jets

Online b -jets were tagged using the $MV2c20$ -tagger^c with two defined efficiency working points: *Tight*, with an overall efficiency of 70% and *Loose* with 85% tagging efficiency.

^bJan 2017 Recommendations: 2016-20_7-13TeV-MC15-CDI-2017-01-31_v1.root

^cMar 2016 Recommendations: 2016-Winter-13TeV-MC15-CDI-March10_v1.root

OBJECT PERFORMANCE

Prior to conducting a full study of TLA on the VBF $H \rightarrow b\bar{b}$ channel, the features of jet objects reconstructed offline and within the HLT were compared to identify any performance differences in the base components of an event reconstruction. The jet objects were compared on a one to one basis, by matching an online jet to an offline jet by requiring the ΔR value between the two jets to be below a threshold value of 0.3^a.

To do: Does this need a plot, or is this sufficient?

4.1 Leading b -jets

The leading p_T offline b -jet selected using the *Tight* working point was matched to a corresponding b -jet using ΔR matching. The following figures show the ratio of the difference in value between the offline and online jet calculated using the following formula for jet feature X :

$$\Delta X_{ratio} = \frac{X_{Offline} - X_{Online}}{X_{Offline}} \quad (4.1)$$

where $X_{Offline}$ is the value of the feature on the offline jet, and X_{Online} is from the HLT jet.

^aDetermined from a plot of ΔR values between all pairs of jets

4.1.1 Monte-Carlo

4.1.1.1 Plots of b -jet features

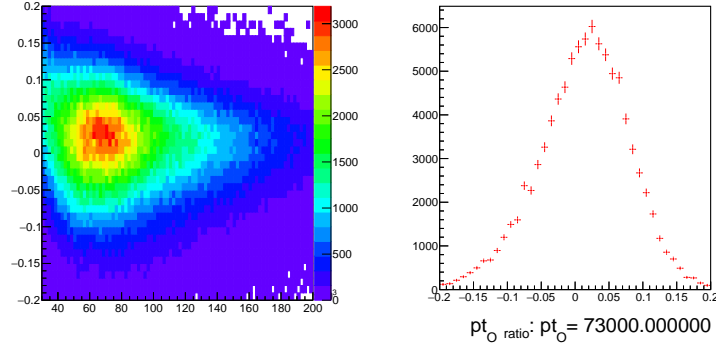


Figure 4.1: $\Delta p_{T \text{ ratio}}$ for the leading p_T b -jet from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79 \text{ GeV}$.

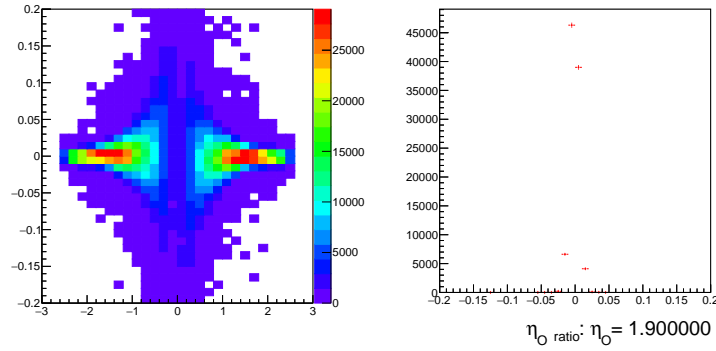


Figure 4.2: $\Delta \eta \text{ ratio}$ for the leading p_T b -jet from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

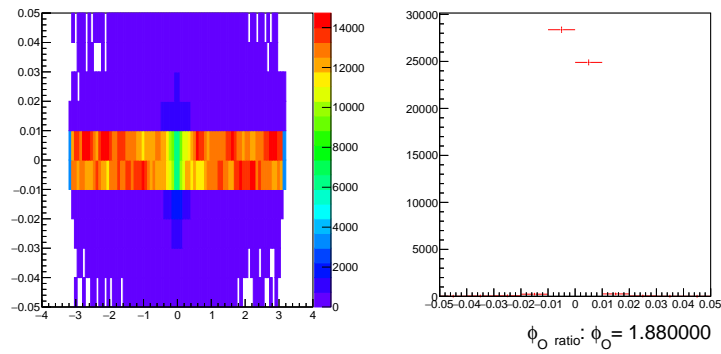


Figure 4.3: $\Delta \phi \text{ ratio}$ for the leading p_T b -jet from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

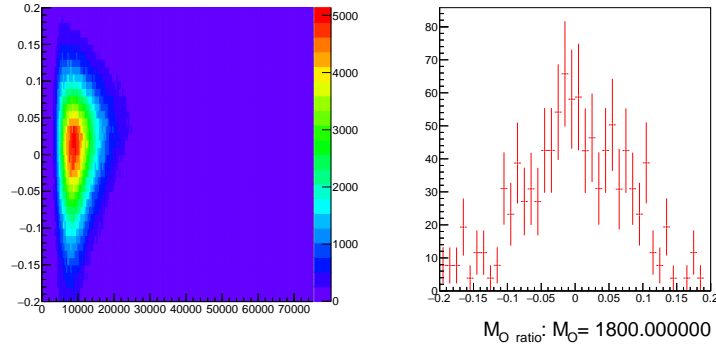


Figure 4.4: ΔM_{ratio} for the leading p_T b -jet from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.1.1.2 Conclusions from MC jet features

4.1.2 Data

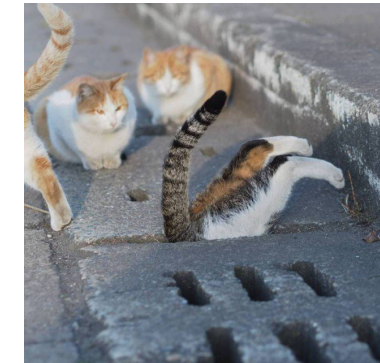
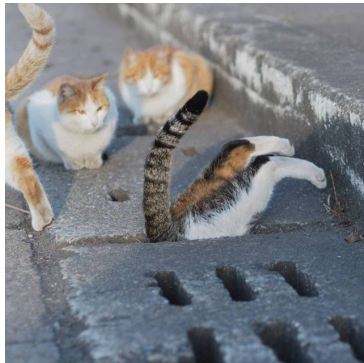


Figure 4.5:

Figure 4.6:

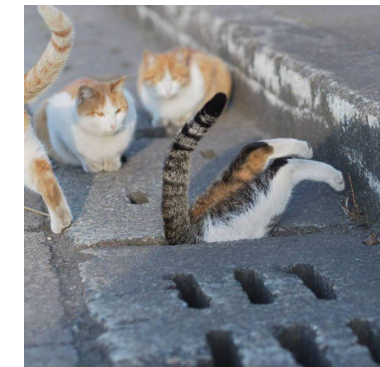
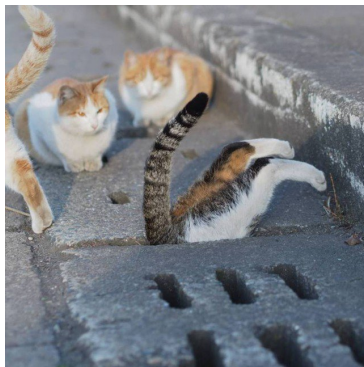


Figure 4.7:

Figure 4.8:

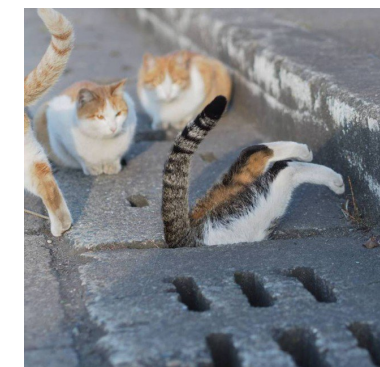
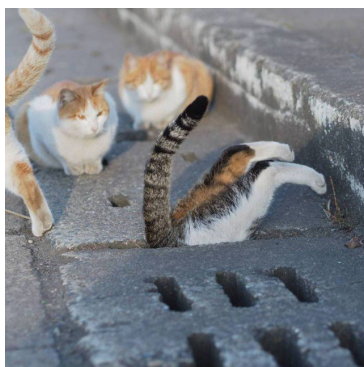


Figure 4.9:

Figure 4.10:



Figure 4.11:

Figure 4.12:

4.2 Leading Non b -jets

The non b -jet category is defined as the jets exclusive to those tagged in Section 4.1. Again, the leading p_T offline jet from this list is matched with an online jet for the comparison.

4.2.1 Monte-Carlo

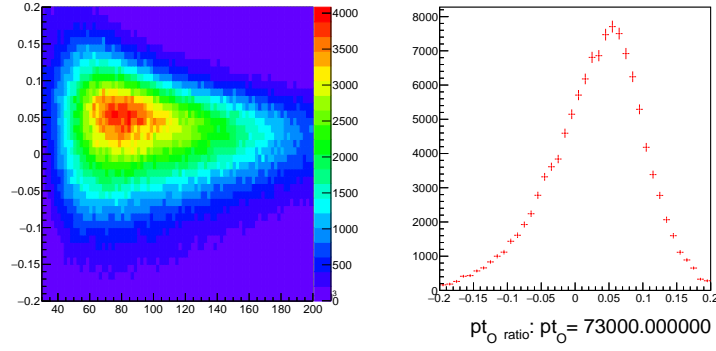


Figure 4.13: Δp_T ratio for the leading p_T non b -jet from MC events against p_T of the offline b -jet.

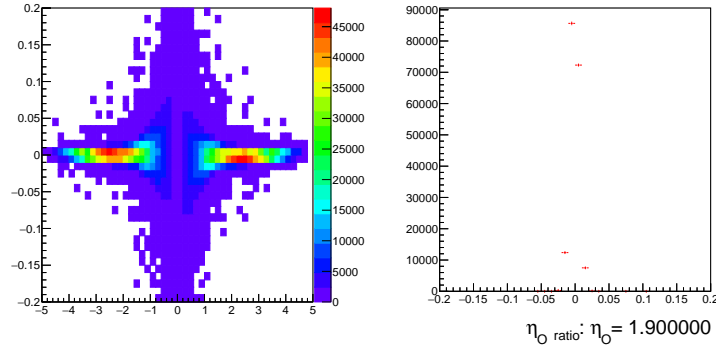


Figure 4.14: $\Delta \eta$ ratio for the leading p_T non b -jet from MC events against η of the offline b -jet.

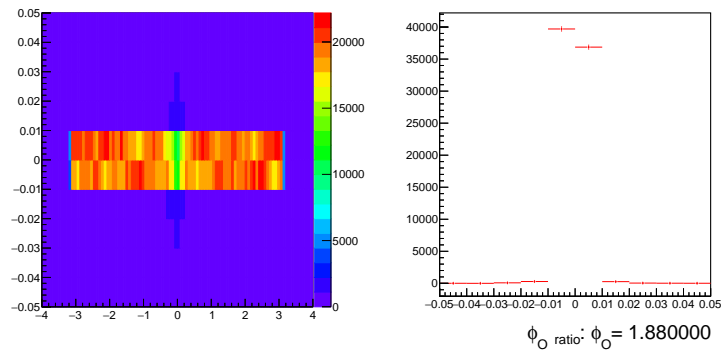


Figure 4.15: $\Delta \phi$ ratio for the leading p_T non b -jet from MC events against ϕ of the offline b -jet.

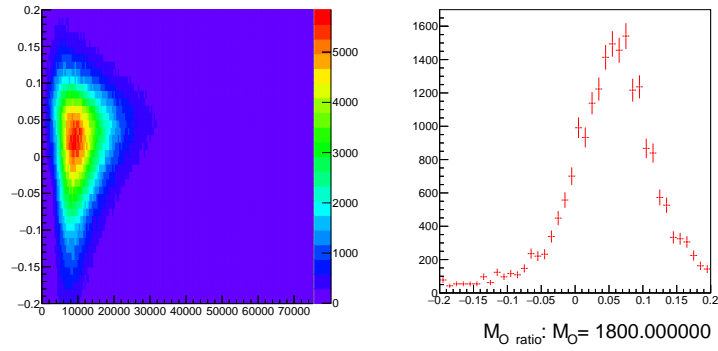


Figure 4.16: ΔM_{ratio} for the leading p_T non b -jet from MC events against M of the offline b -jet.

4.2.2 Data



Figure 4.17:

Figure 4.18:

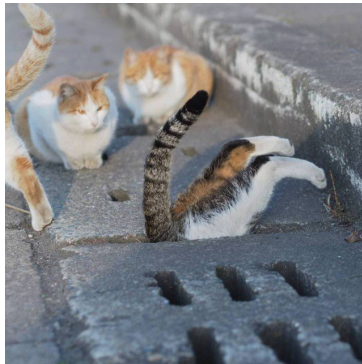


Figure 4.19:

Figure 4.20:

4.3 Central Jets

4.3.1 Monte-Carlo

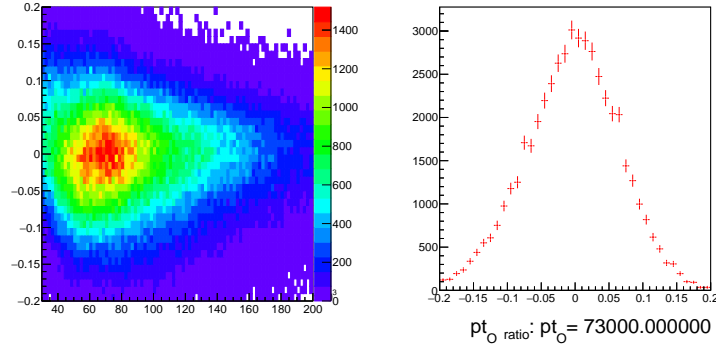


Figure 4.21: $\Delta p_T \text{ ratio}$ for the leading p_T b -jet with $0 < \eta < 1$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

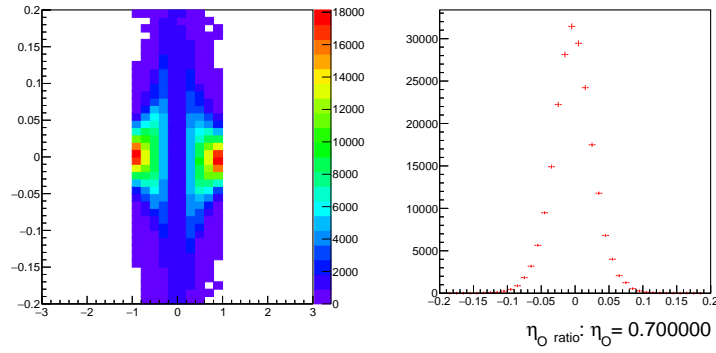


Figure 4.22: $\Delta \eta \text{ ratio}$ for the leading p_T b -jet with $0 < \eta < 1$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

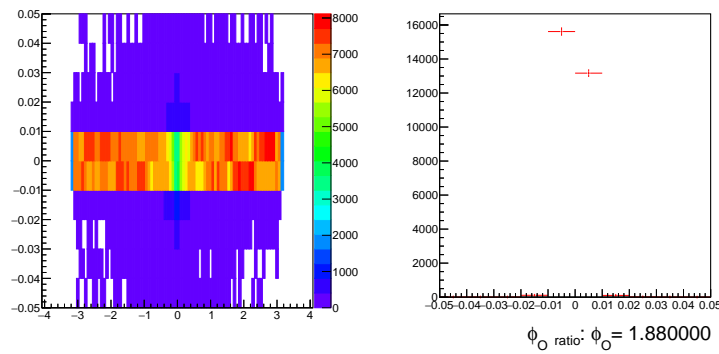


Figure 4.23: $\Delta \phi \text{ ratio}$ for the leading p_T b -jet with $0 < \eta < 1$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

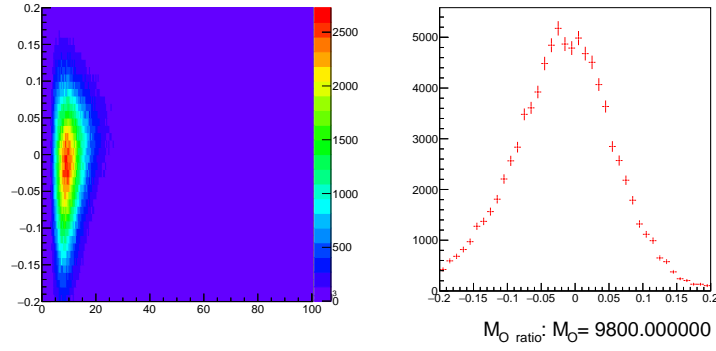


Figure 4.24: ΔM_{ratio} for the leading p_T b -jet with $0 < \eta < 1$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.3.2 Data

4.3.3 Monte-Carlo

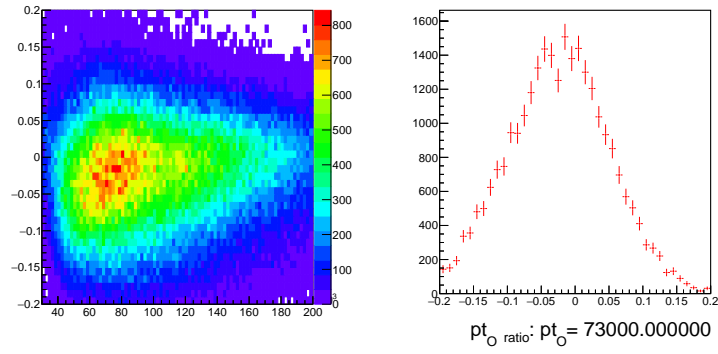


Figure 4.25: Δp_T_{ratio} for the leading p_T non b -jet with $0 < \eta < 1$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

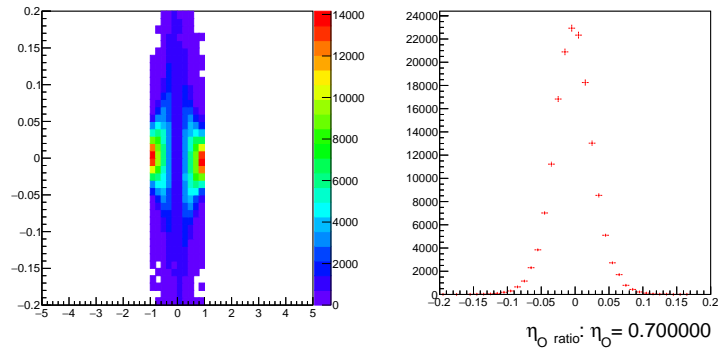


Figure 4.26: $\Delta \eta_{ratio}$ for the leading p_T non b -jet with $0 < \eta < 1$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

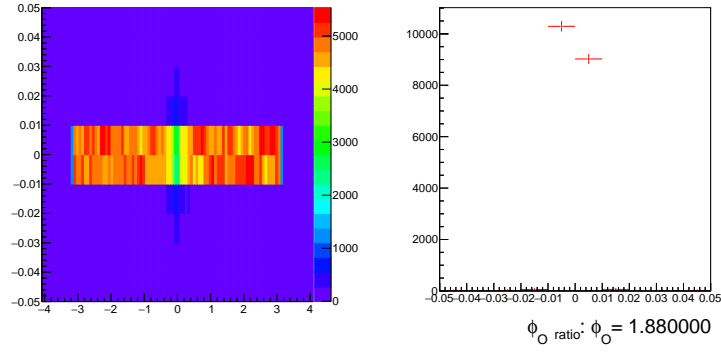


Figure 4.27: $\Delta\phi_{ratio}$ for the leading p_T non b -jet with $0 < \eta < 1$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

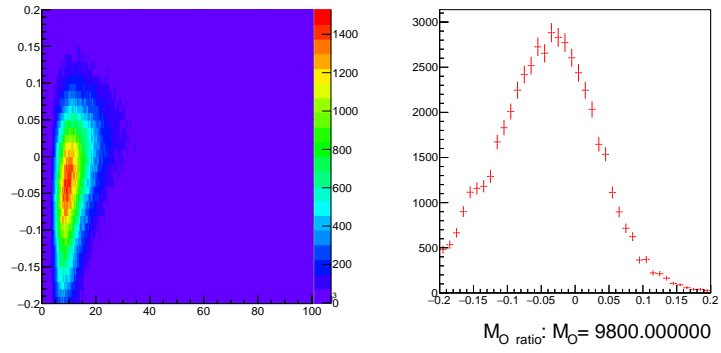


Figure 4.28: ΔM_{ratio} for the leading p_T non b -jet with $0 < \eta < 1$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.3.4 Data

4.4 Core

4.4.1 Monte-Carlo

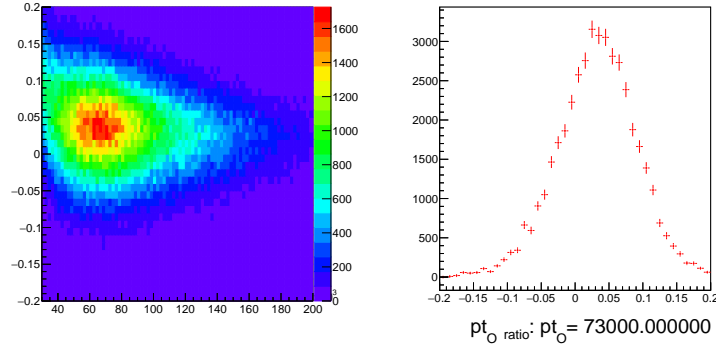


Figure 4.29: Δp_T ratio for the leading p_T b -jet with $1 < \eta < 2.4$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

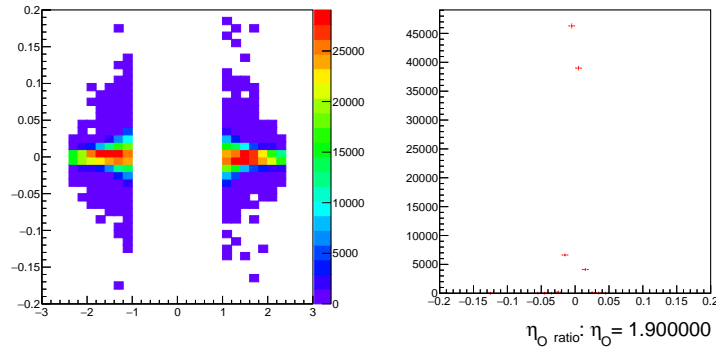


Figure 4.30: $\Delta \eta$ ratio for the leading p_T b -jet with $1 < \eta < 2.4$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

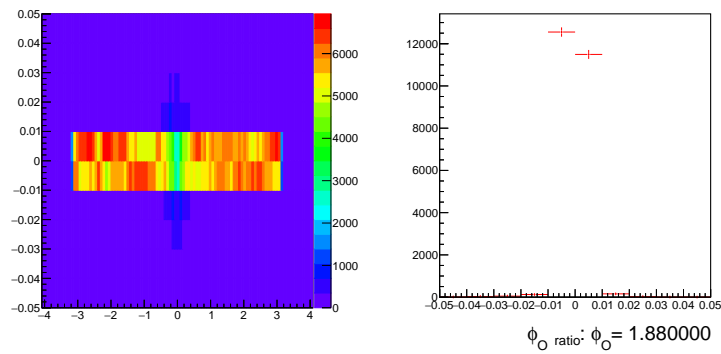


Figure 4.31: $\Delta \phi$ ratio for the leading p_T b -jet with $1 < \eta < 2.4$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

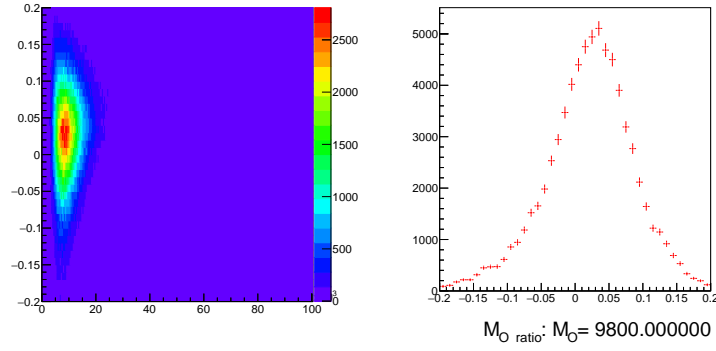


Figure 4.32: ΔM_{ratio} for the leading p_T b -jet with $1 < \eta < 2.4$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.4.1.1 Data

4.4.2 Non Bjets

4.4.3 Monte-Carlo

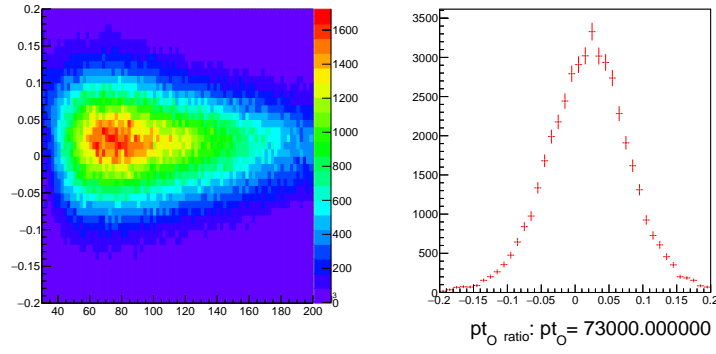


Figure 4.33: $\Delta p_T ratio$ for the leading p_T non b -jet with $1 < \eta < 2.4$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

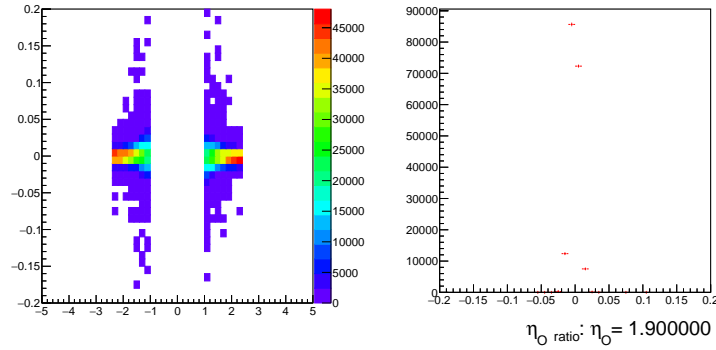


Figure 4.34: $\Delta \eta_{ratio}$ for the leading p_T non b -jet with $1 < \eta < 2.4$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

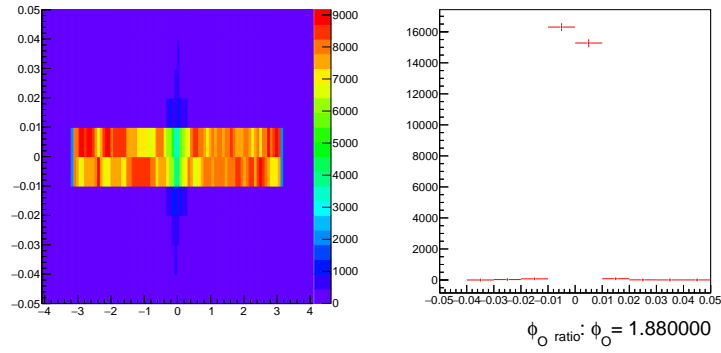


Figure 4.35: $\Delta\phi_{ratio}$ for the leading p_T non b -jet with $1 < \eta < 2.4$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

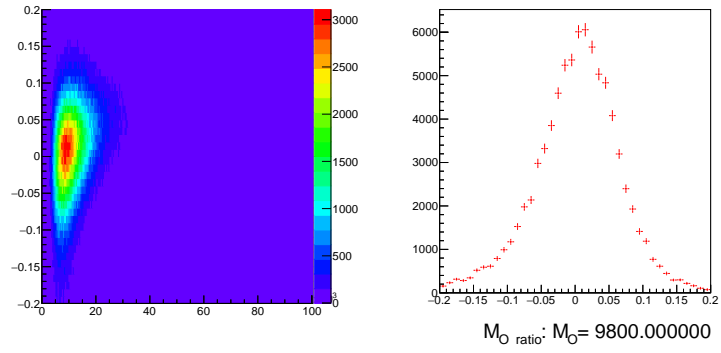


Figure 4.36: ΔM_{ratio} for the leading p_T non b -jet with $1 < \eta < 2.4$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.5 Forward Jets

4.5.1 Monte-Carlo

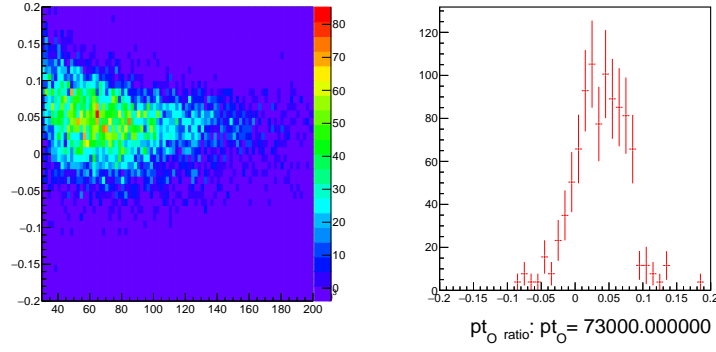


Figure 4.37: $\Delta p_T \text{ ratio}$ for the leading p_T b -jet with $2.4 < |\eta|$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

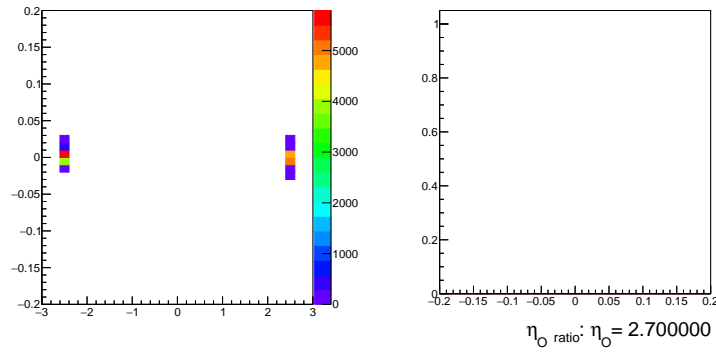


Figure 4.38: $\Delta \eta \text{ ratio}$ for the leading p_T b -jet $2.4 < |\eta|$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

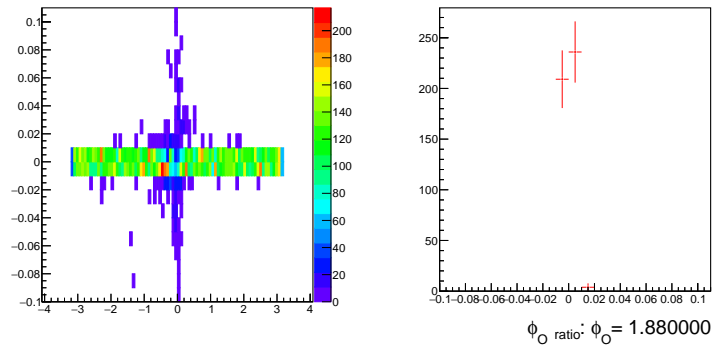


Figure 4.39: $\Delta \phi \text{ ratio}$ for the leading p_T b -jet $2.4 < |\eta|$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

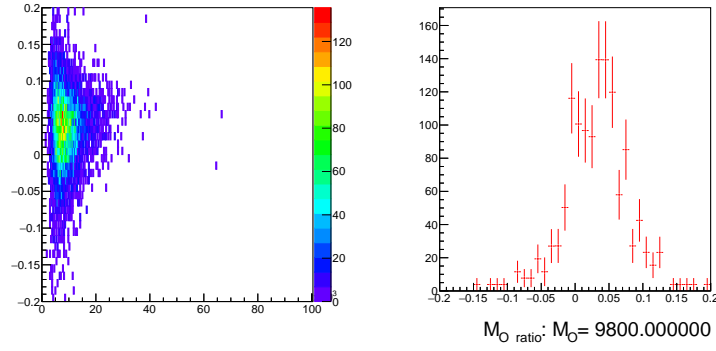


Figure 4.40: ΔM_{ratio} for the leading p_T b -jet $2.4 < |\eta|$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.5.2 Non bjets

4.5.3 Monte-Carlo

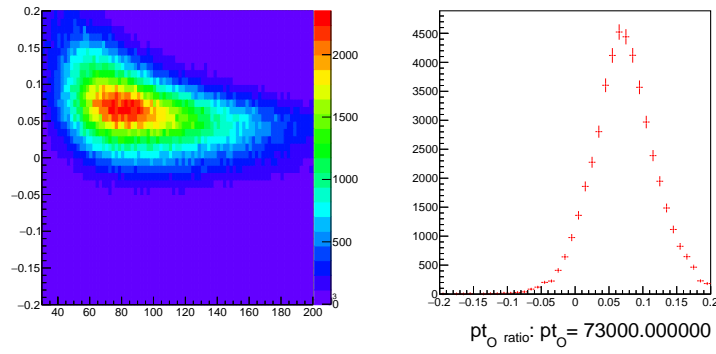


Figure 4.41: Δp_T_{ratio} for the leading p_T non b -jet $2.4 < |\eta|$ from MC events against p_T of the offline b -jet. A slice across the y -axis has been taken at $p_T = 79\text{GeV}$.

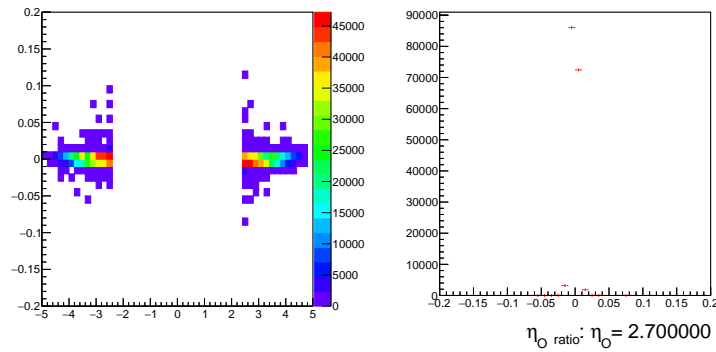


Figure 4.42: $\Delta \eta_{ratio}$ for the leading p_T non b -jet $2.4 < |\eta|$ from MC events against η of the offline b -jet. A slice across the y -axis has been taken at $\eta = -1.9$.

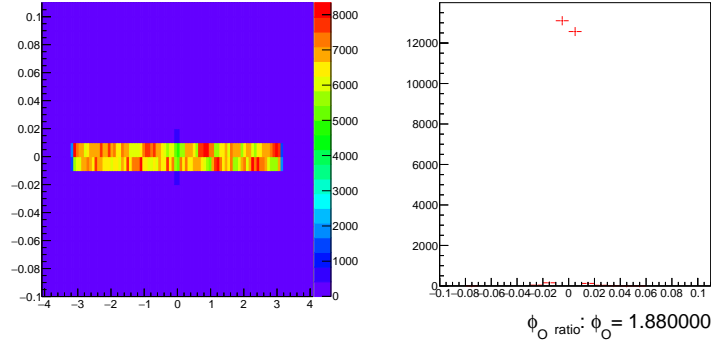


Figure 4.43: $\Delta\phi_{ratio}$ for the leading p_T non b -jet $2.4 < |\eta|$ from MC events against ϕ of the offline b -jet. A slice across the y -axis has been taken at $\phi = -1.64$.

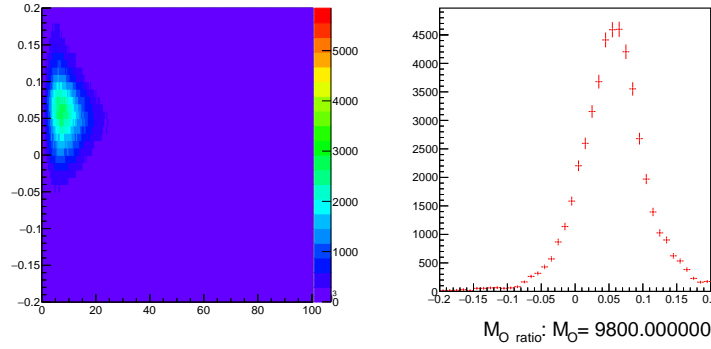


Figure 4.44: ΔM_{ratio} for the leading p_T non b -jet $2.4 < |\eta|$ from MC events against M of the offline b -jet. A slice across the y -axis has been taken at $M = 7\text{GeV}$.

4.6 Jet Tagging Efficiency

As covered in 2.5.1, the standard algorithm for 2016 physics analyses was chosen to be the 2016 MV2c10 algorithm. However, the HLT b -tagging algorithm uses the MV2c20 algorithm. [5] To perform a valid TLA the performance of the tagging algorithms between trigger level and offline must be similar. With the datasets used for this analysis (??) the MC data produced in 2015 would make use of the older configurations compared to the newer configurations in the data.

Here the tagging efficiency of the HLT and offline taggers is studied for different jet flavours in the MC sample. An offline/HLT jet pair was formed using ΔR matching and truth label of the jet used to assign a flavour. The efficiency plots in figures 4.45, 4.46 and 4.47 show the fraction of these jets that were identified as b -jets by the HLT and offline tagging algorithms.

not sure what effect
this config changes
had on the trigger

4.6.1 b -jet efficiency

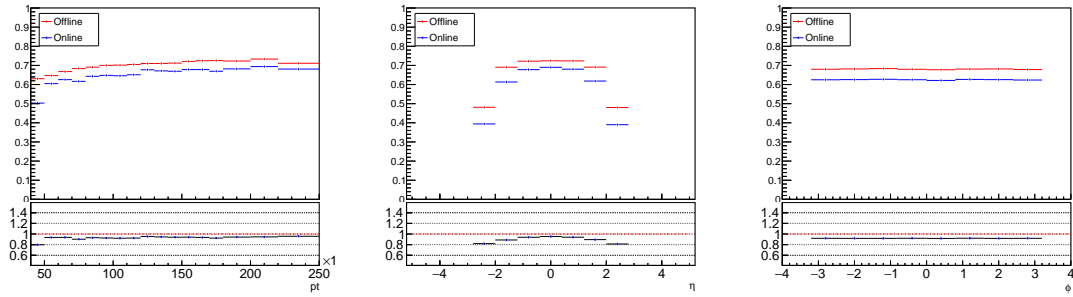


Figure 4.45:

4.6.2 c -jet efficiency

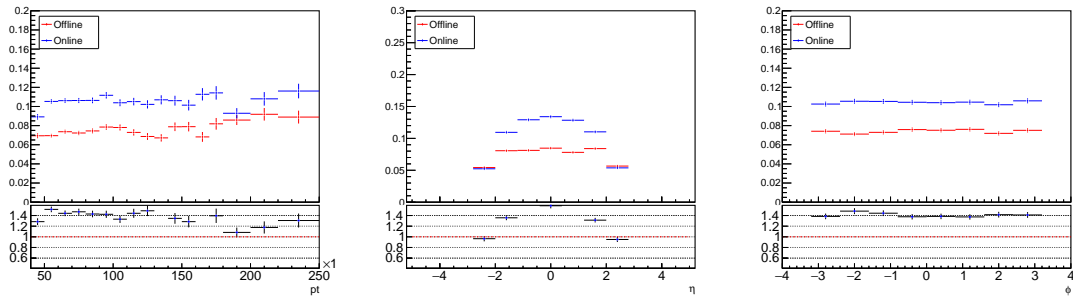


Figure 4.46:

To do: Options, could show more vars or alternatively the reference hists, or alternatively just reference the references

4.6.3 Light-jet efficiency

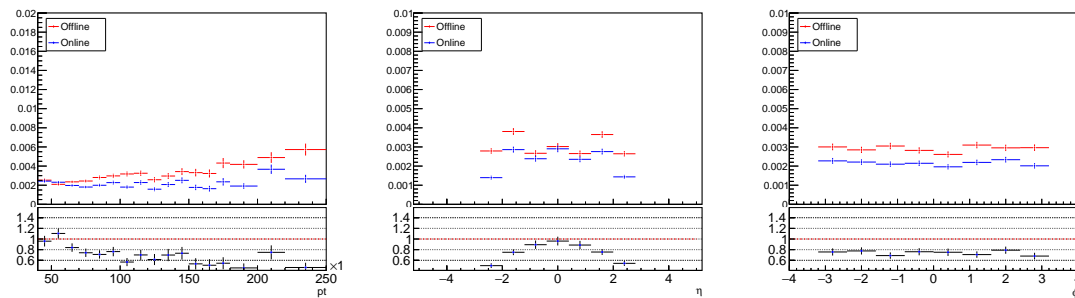


Figure 4.47:

4.6.4 Tag Matching

For each pair of jets that could be matched between online and offline, and then successfully have a b -tagging decision evaluated on the jets, the agreement of the b -tagging between the two jets was checked. These were found to match one another in 90.91% of cases.

4.6.5 Comparison of HLT and offline tagging efficiencies

Primarily considering the p_T plots of efficiency, the HLT b -tagging is found to be around 5% less efficient than the offline b -tagging for jets with $p_T > 50$ GeV. This is a consistent direction of efficiency shift as found when comparing the 2016 MV2c10 and 2015 MV2c20 algorithms on the training $t\bar{t}$ sample, but of a larger magnitude. The increase in the rate of c -jet mistagging is absolutely consistent with the refinements to the algorithm between the 2016 MV2c10 and 2015 MV2c20, with increased levels of c -jet rejection in the offline 2016 MV2c10, and the $\sim 40\%$ increase is consistent with the expected shift from the optimised algorithm. [3] The light-jet behaviour is also similar as expected but ????

To do: some light
jet related
shenanigans

4.7 MV2 Discriminant Values - ???

To do: Necessary

Here would show
plots of the MV2
value against p_T/η
or whatever

4.8 MV2 Input Variables - ???

KINEMATICS

5.1 Specific Jet Feature Distributions

For the standard set of jet features, plot the overall distributions in a standard 2 hist ratio plot for data and MC. This also includes jet counts possibly

5.2 Specific Jet Feature Distributions

5.2.1 Two Central Channel

5.2.1.1 p_T

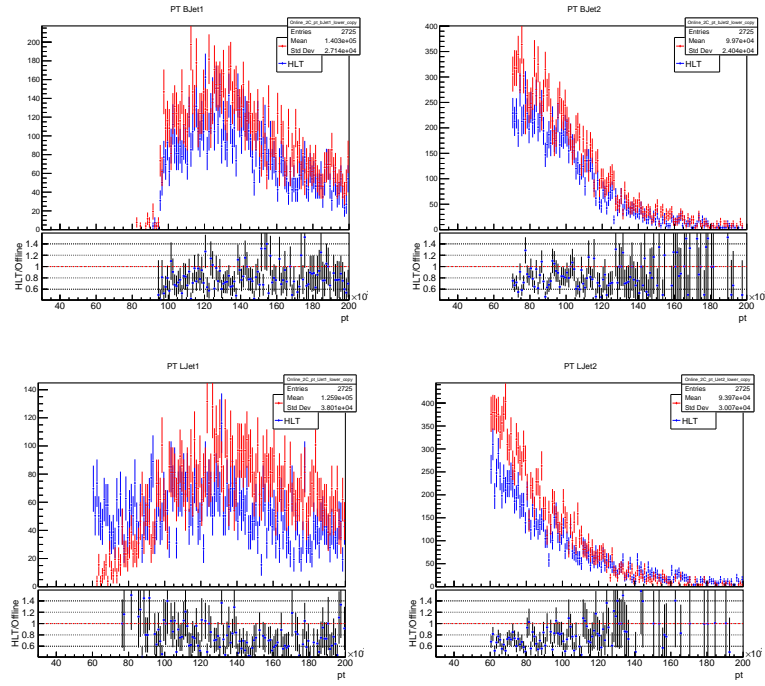
5.2.1.2 η

5.2.1.3 ϕ

5.2.1.4 M

5.3 BDT Input Variables

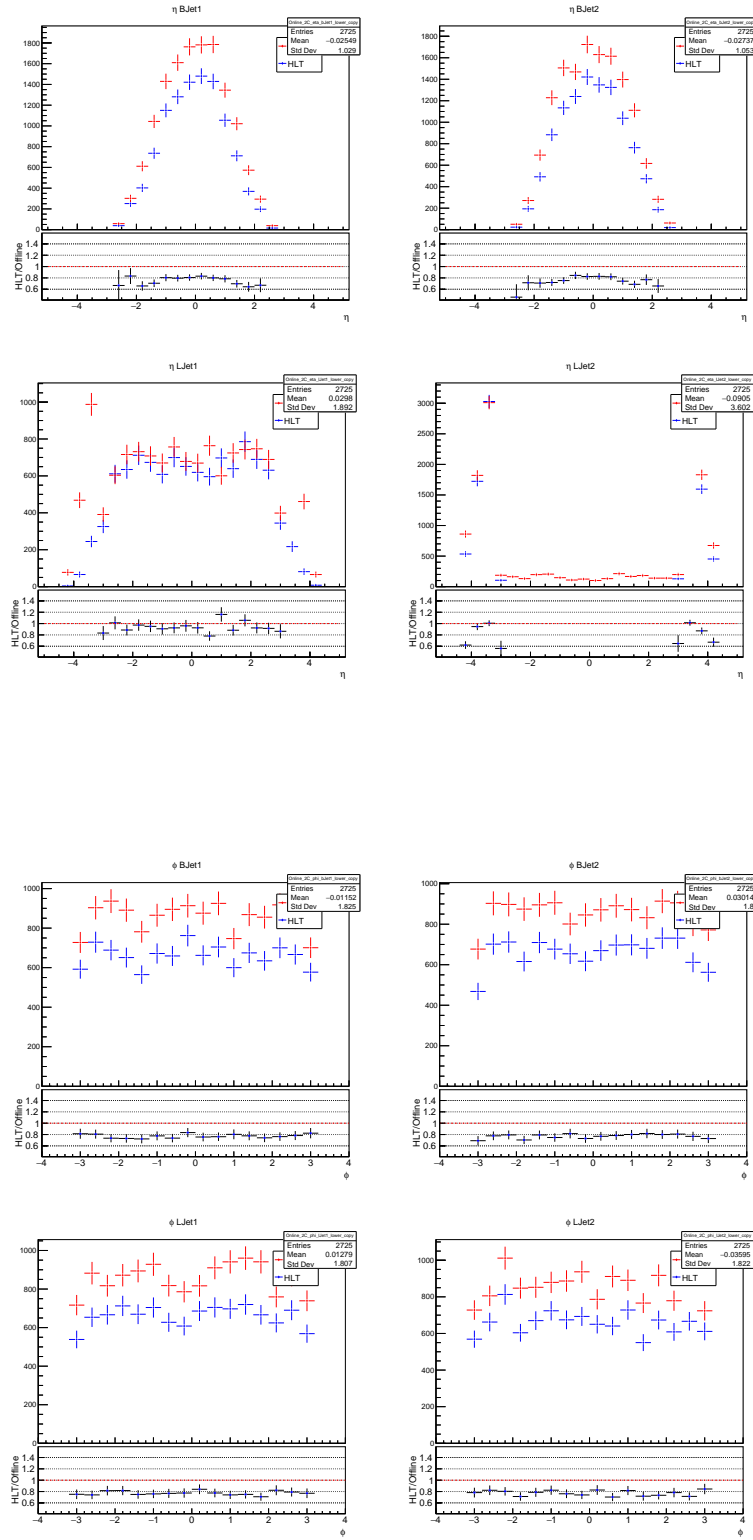
- M_{jj}
- $p_{T,jj}$
- $\cos \theta$
- $\Delta\eta_{jj}$
- $Max(\eta)$
- η^*



- $\min \Delta R(j_1)$
- $\min \Delta R(j_1)$
- p_T balance
- $N_{TRK}(j_1)PV500$?
- $N_{TRK}(j)PV500$?

5.4 Mbb Distribution

Prior paper suggests this is the 'final' plot, a shape comparison between BDT influenced control and signal regions of the Mbb distribution. A little confused as to exactly what we need here.



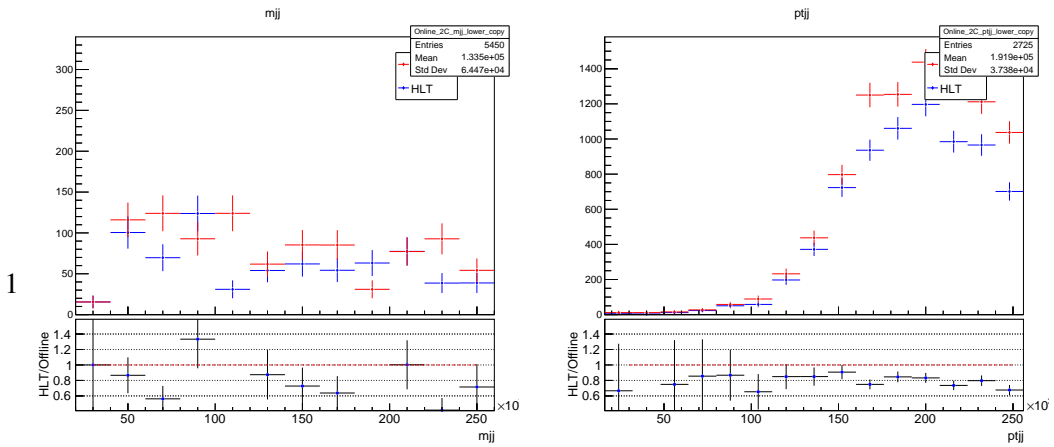
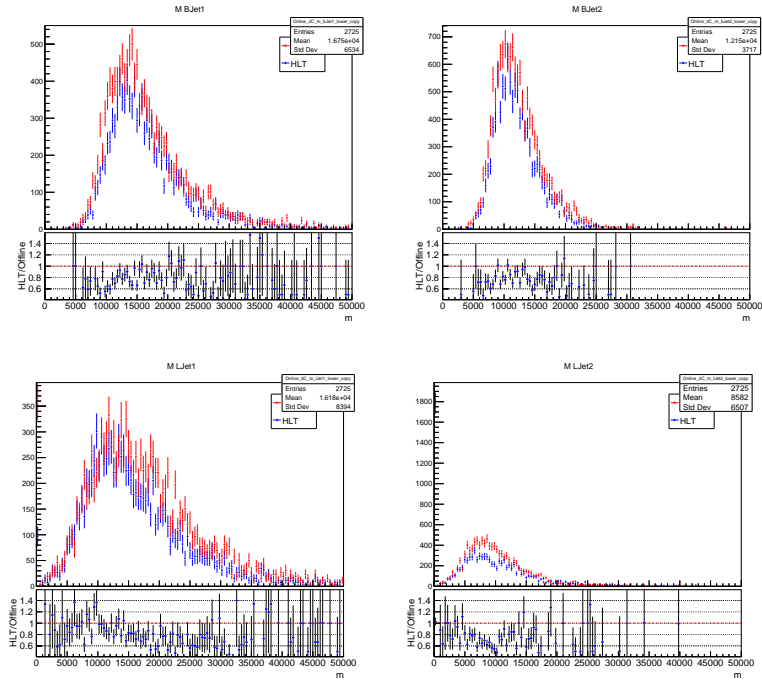


Figure 5.1:

Figure 5.2:

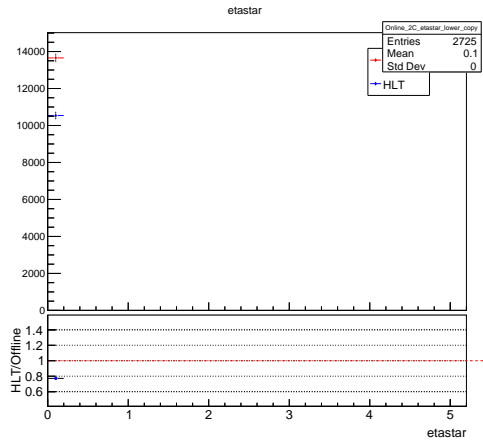


Figure 5.3:

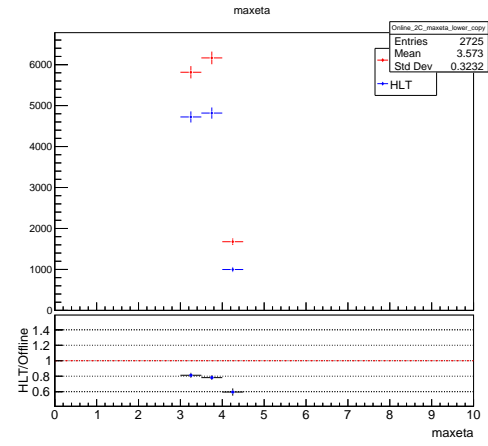


Figure 5.4:

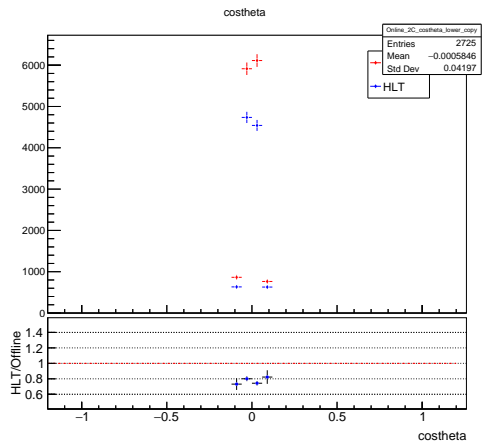


Figure 5.5:

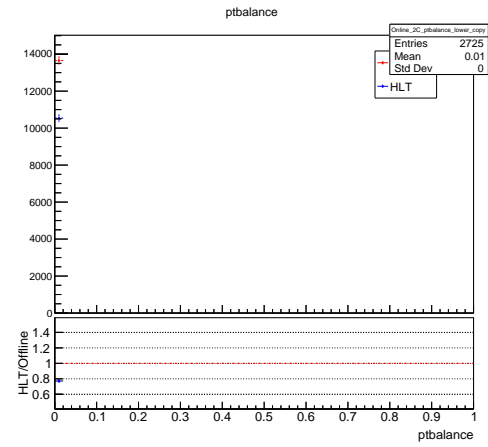


Figure 5.6:

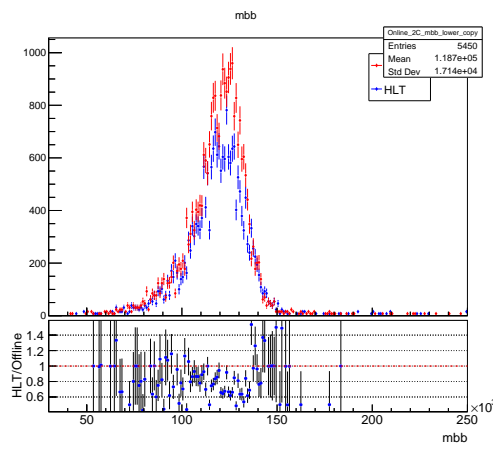


Figure 5.7:

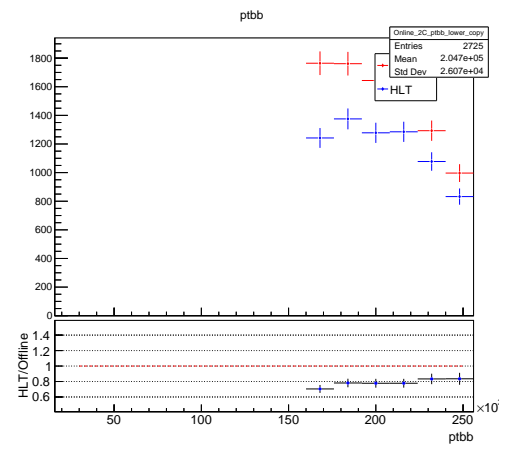


Figure 5.8:

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- [1] LHC Higgs Cross Section Working Group Collaboration, S. Dittmaier et al., *Handbook of LHC Higgs Cross Sections: 1. Inclusive Observables*, [arXiv:1101.0593 \[hep-ph\]](#).
- [2] S. Asai et al., *Prospects for the search for a standard model Higgs boson in ATLAS using vector boson fusion*, *Eur. Phys. J. C* **32S2** (2004) 19–54, [arXiv:hep-ph/0402254 \[hep-ph\]](#).
- [3] ATLAS Collaboration, *Optimisation of the ATLAS b-tagging performance for the 2016 LHC Run*, ATL-PHYS-PUB-2016-012 (2016). <https://cds.cern.ch/record/2160731>.
- [4] ATLAS Collaboration, *Performance of b-Jet Identification in the ATLAS Experiment*, *JINST* **11** no. 04, (2016) P04008, [arXiv:1512.01094 \[hep-ex\]](#).
- [5] ATLAS Collaboration, M. Aaboud et al., *Performance of the ATLAS Trigger System in 2015*, *Eur. Phys. J. C* **77** no. 5, (2017) 317, [arXiv:1611.09661 \[hep-ex\]](#).