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COLLEGE OF ARTS AND SCIENCES

SEARCH FOR HIGGS BOSON DECAYS TO LEPTOPHILIC LONG-LIVED SCALAR
PARTICLES WITH REGIONS OF INTEREST (ROIS) AND MACHINE LEARNING IN CMS

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To my parents, who always suspected I'd end up here

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Thanks to many people.

TABLE OF CONTENTS

| | |
|---|-----------|
| List of Tables | vii |
| List of Figures | viii |
| Abstract | x |
| 1 Introduction | 1 |
| 2 The Standard Model and Its Future | 4 |
| 3 The CMS Detector | 5 |
| 3.1 The LHC and the CMS | 5 |
| 3.2 Tracker of the CMS | 5 |
| 3.3 HGCAL of the CMS | 5 |
| 4 Trigger Strategy | 6 |
| 4.1 Global Tags | 6 |
| 4.2 Trigger Paths | 6 |
| 5 Physics Object Definitions | 9 |
| 5.1 Muons | 9 |
| 5.2 Jets | 10 |
| 5.3 Taus | 11 |
| 5.4 Region of Interest | 11 |
| 5.4.1 Tracks | 12 |
| 5.4.2 Vertex Fitter | 12 |
| 5.4.3 ROI formation | 12 |
| 5.5 Pileup Reweighting | 13 |
| 6 Machine Learning | 17 |
| 6.1 Machine Learning Software | 17 |
| 6.2 Machine Learning Input Variable | 18 |
| 6.3 DNN Variable Test | 18 |
| 6.4 Signal and Background MC Test | 19 |

| | | |
|-------------------------------|--|-----------|
| 6.5 | SHAP values | 20 |
| 7 | Event Selection, Signal and Control Regions | 22 |
| 7.1 | Delta Phi(lead ROI,sublead ROI) | 23 |
| 7.2 | Number of Annulus Tracks Associated with ROI | 24 |
| 7.3 | Isolation criteria for muons | 25 |
| 7.4 | Leading muon's transverse impact parameter to PV | 25 |
| 7.5 | DeltaR(ROI, jet) | 26 |
| 8 | Background Estimation | 28 |
| 9 | Systematic Uncertainties | 30 |
| 10 | Results | 31 |
| 11 | Conclusions | 32 |
| Appendix | | |
| A | Data Samples | 33 |
| A.1 | Monte Carlo Samples | 34 |
| A.1.1 | Signal Model and Simulation | 34 |
| A.1.2 | Background Monte Carlo | 34 |
| B | Others | 38 |
| References | | 39 |
| Biographical Sketch | | 44 |

LIST OF TABLES

| | | |
|-----|---|----|
| 4.1 | Data and MC Global tags used 2018 | 6 |
| 4.2 | HLT trigger paths used in the analysis 2018. | 7 |
| 6.1 | Tensorflow information | 18 |
| 6.2 | ROI (trackCluster) variables by category | 18 |
| 6.3 | ROI (Annulus) variables by category | 19 |
| 6.4 | Event variables by category | 19 |
| A.1 | Datasets used in the analysis | 33 |
| A.2 | QCD MuEnriched Pt5 background Monte Carlo samples | 36 |
| A.3 | W,Z,H boson background Monte Carlo samples | 36 |
| A.4 | Top background Monte Carlo samples | 37 |
| A.5 | Monte Carlo sample summary | 37 |

LIST OF FIGURES

| | | |
|-----|--|----|
| 4.1 | asd | 7 |
| 4.2 | eeee | 8 |
| 5.1 | Data/MC of muon objects | 9 |
| 5.2 | Data/MC of jet objects | 10 |
| 5.3 | Data/MC of tau objects | 11 |
| 5.4 | Data/MC of ROI distribution | 13 |
| 5.5 | 2Data/MC of ROI distribution | 13 |
| 5.6 | ww | 14 |
| 5.7 | dd | 15 |
| 5.8 | ed | 15 |
| 5.9 | aaaa | 16 |
| 6.1 | asdasdasd | 19 |
| 6.2 | Data/MC agreement for ROI scores | 20 |
| 6.3 | Data/MC agreement for loglead/sublead scroes | 21 |
| 7.1 | Signal versus Background for $\log_{10}(1-\text{ROIscore})$, where the ROI score is the highest ROI of the event. Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 22 |
| 7.2 | Signal versus Background for $\log_{10}(1-\text{subROIscore})$, where the ROI score is the second highest ROI (outside of $d\Phi=0.4$ from leading ROI) of the event. Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 23 |
| 7.3 | Signal versus Background for Delta Phi(leadROI, subleadROI). Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 24 |
| 7.4 | Number of tracks in the annulus cone of the leading ROI. Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 25 |
| 7.5 | leading muon's transverse impact parameter value to the primary vertex. Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 26 |
| 7.6 | Delta R(Jet, leadingROI). Left plot is for MS-15.ctauS-10mm point, whereas the right plot is for MS-15.cauS-100mm point | 27 |

| | | |
|------|---|----|
| 8.1 | Cutflow histogram of MS15GeV-ct100mm point. Left plot is for region A, whereas the right plot is for region D | 29 |
| 8.2 | eee | 29 |
| 10.1 | Current Preliminary limit plots | 31 |
| 10.2 | Current Preliminary limit plots | 31 |
| A.1 | Leading Feynman diagrams for ggH production mode | 34 |
| A.2 | pt of the scalar products | 35 |
| A.3 | DeltaR of the scalar products | 35 |
| A.4 | lifetime of the scalar products in the lab frame | 36 |

ABSTRACT

We present a search for long-lived particles (LLPs) produced in gluon fusion Higgs production mode (ggH), using a novel strategy of Regions of Interest (ROIs). Regions of Interest are collections of pair-wise track vertices fitted by the vertex fitter in CMSSW. The analysis focuses LLPs with lifetimes that result in decays in the tracker region, with concentration on the ggH production mode for the highest Higgs cross-section. Variables of the constructed ROIs become inputs for our Deep Neural Network (DNN) Machine Learning (ML) algorithms, as a main discriminator between the signal and the background. We focus on the Standard Model (SM) τ lepton final state. This final state is particularly interesting, given τ lepton final state exclusion limits are mostly omitted in precedent analyses, due to τ leptons' non trivial reconstruction mechanisms in CMS. In order to trigger on the ggH production mode of the Higgs particle, decaying into its leptonic signature via exotic long-lived scalar particles, we exploit the B-parking trigger. B-parking trigger is newly installed High Level Trigger path in the CMS detector, targetting soft displaced muon. The trigger was installed in 2018, totalling in integrated Luminosity value of 41fb^{-1} . No excess of events over the standard model expectation is observed. The results are interpreted in the context of exotic Higgs decays to a pair of long-lived scalars (S). We set limits on the branching ratio of the Higgs to LLPs, $\mathbf{B}(\text{H} \rightarrow SS \rightarrow \tau\tau\tau\tau)$, as a function of the proper lifetime. The analysis has strongest discriminant power for the scalar particle's lifetime from 1mm to 10mm for 7 and 15 GeV, and from 10mm to 100mm for 40 and 55 GeV. The analysis' exclusion limit on the branching ratio for τ lepton final state ($\mathbf{B}(\text{H} \rightarrow SS \rightarrow \tau\tau\tau\tau)$) is one of the most stringent results from the LHC detectors. The 9% at 10mm for 7 GeV, 5% at 10mm for 15 GeV, 9% at 100mm for 40 GeV, and 20% at 1000mm for 55 GeV are respective values.

CHAPTER 1

INTRODUCTION

The discovery of particles at the electroweak scale, such as the top quark at Fermilab's CDF and D0 [1, 2] and the Higgs boson at the Large Hadron Collider (LHC) in CERN [3, 4], led to discovery of all constituents in the Standard Model (SM). The SM describes the nature of fundamental particles and their interactions with precision. In spite of its success, the SM suffers from the few obstacles: the evidence of neutrino masses and mixing [5], the observations of bullet clusters confirming the presence of dark matter (DM) [6, 7, 8, 9, 10], and baryon-antibaryon asymmetry [11] all remain unexplained in the framework of SM. In addition, the SM suffers from the naturalness problem. To solve such issues, one needs to look for physics Beyond the Standard Model (BSM).

The naturalness problem originates from the fact that the SM Higgs is a scalar particle. Unlike fermions or gauge bosons, its mass is not protected by any symmetry and is subject to large radiative corrections, especially from the top quark loop. Thus, for the SM to be valid up to the Planck or Grand Unification Theory (GUT) scales, the necessary radiative corrections are enormous. One needs an exorbitant amount of fine-tuning to fit the Higgs mass at the observed value of 125GeV. One of the most popular solutions to this problem is Supersymmetry (SUSY), which assigns chirality to the Higgs particle. SUSY solves the fine-tuning problem, neutrino masses, and provides a candidate for DM. Unfortunately, the LHC has found no significant excess over the SM background in their search for SUSY[12]. Although the non-observation of supersymmetric partner particles does not invalidate SUSY, it makes less attractive among the particle physics community. Non-observation of superpartners, particularly the stop (scalar partner of the top quark) has pushed its mass beyond 1TeV. This generates "little hierarchy" problem, but an alternative solution of "neutral naturalness" remains.

In the framework of neutral naturalness, the top partners are not charged under the SM color group. Because of being colorless, their production crosssection is much smaller, and the present limits on the top partner particles are well below 1TeV. Examples of neutral naturalness models are the Twin Higgs [13], Folded SUSY [14], and the Quirky Little Higgs [15] models. Theoretical models provide the possibility of neutral Long-Lived Particles (LLPs), which may be produced in the proton-proton collisions of the LHC, and decay back to SM particles far from the interaction

point (IP).[16] If the mirror QCD gluons form scalar glueballs, the SM Higgs boson can become a "Higgs Portal" between the SM and BSM mirror QCD scalar glueballs. In the Mirror SM and Twin SM models, only the SM Higgs boson can interact with both SM QCD and mirror QCD particles. BSM mirror QCD scalar glueballs can only decay back to SM particles via Higgs boson decay as well. Because of its decay as an offshell Higgs boson, its crosssection is highly suppressed. Decay branching ratio to highest mass fermions will be highest following the Yukawa couplings. Decay ratio into b quarks or tau leptons are highest depending on the mirror scalar's mass. The displaced decays of the scalars would lead to exotic signatures in the CMS, such as distant innermost tracker hit, displaced vertices, and displaced jets. Phenomenology of long-lived particles in LHC entailed increase in interest of neutral naturalness framework among the particle physics community. [17, 18]. The long-lived scalar model is shown in the Figure (to be made soon).

Searches for LLPs decaying into final states containing jets were investigated at the Tevatron ($\sqrt{s} = 1.96$ TeV) by both CDF [19] and D0 [20] Collaborations, at the LHC by the ATLAS and LHCb Collaborations at $\sqrt{s} = 7$ TeV [21, 22], by the ATLAS, CMS and LHCb Collaborations at $\sqrt{s} = 8$ TeV [23, 24, 25, 26, 27, 28, 29]. More recently, by the CMS [30, 31, 32, 33, 34, 35] and ATLAS Collaborations [36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47] at $\sqrt{s} = 13$ TeV. CMS Collaboration released a new result in 2021, in which the Higgs is created in association with a Z vector boson [48], for a better probe into lighter scalar masses thanks to its clean dilepton trigger.

Although exclusion limit on branching ratio of the Higgs to the LLPs to b and d-quark were set below unitarity for analyses above, exclusion limit for τ final state has been frequently omitted or presented with values above 1. Displaced Jets analyses face challenges for τ final state, due to τ 's non-trivial hadronic and leptonic decay modes and complicated reconstruction mechanism. However, the Leptophilic model for Twin Higgs and other Higgs models are also highly motivated [49]. Continuous neglect of τ final state limit is not only a poor practice, but overlooks an important unexplored phase space. This analysis searches for Higgs Portal model with the Higgs' Leptophilic nature, which focuses on τ final state. The 55 GeV maximum scalar mass is set to investigate only on-shell neutral scalar particles from the Higgs. A minimum 7 GeV mass for scalar particles is required to create on-shell tau-lepton pairs. Feynman diagram of the scalar particle production mechanism is depicted in Figure (to be made soon).

Most CMS searches are not optimal for detecting Higgs boson decays to displaced jets due to the soft p_T nature of its decays products. Low hadronic activity in transverse signature becomes particularly more difficult with a long-lived signature. Higgs produced in association with Z vector

boson analysis [48] overcame this barrier with help of dilepton trigger. Although ggH production mode gives the largest Higgs crosssection, it complicates the trigger strategy even further. This analysis exploits the τ lepton’s leptonic decay, in which the τ lepton decays into a soft muon, using a trigger of B Parking High Level Trigger (HLT) Path implemented in CMS for the 2018 portion of Run 2.

Another challenge for a τ lepton analysis is on the different decay modes of τ leptons. τ leptons decay hadronically and leptonically, with several different sub-decay modes. Developing analysis strategies to optimize the search for each sub-decay modes is extremely complicated, a main reason for the omission or no good exclusion limit in precedent LHC results. To be inclusive of all τ leptons’ decay modes, a displaced vertex search can be more efficient than a displaced objects (jet, muon, electrons) search. We exploit the newly developed Regions of Interest mechanism in the tracker volume. Regions of Interest (ROI) form displaced vertex candidates, by fitting pair-wise tracks of Lost-tracks and PackedPFCandidates classes in MINIAOD data into a vertex. ROIs save all relevant track and fitted vertex qualities along with isolation information. These variables are used as input for Machine Learning (ML) algorithms, enabling a highly generic and data-scientific search method.

The rest of the dissertation is organized as follows. In Section 2, we discuss the theoretical background for the need of the BSM theory in more details. In Section 3, the CMS detector is thoroughly discussed with emphasis on the tracks and the calorimeter, which are relevant detector parts for the analysis. We discuss how the analysis exploited the b-parking trigger in Section 4, with description of its original motive for the trigger’s implementation. The physics objects and formation of Regions of Interest are described in Section 5. The machine learning algorithms are further explained in Section 6. The event selections are presented in Section 7. Section 8 describes the data driven background estimation method. Section 9 describes the background estimation method’s validation process and systematic uncertainties. Finally, Section 10 presents the results of the search. We conclude with Section 11.

CHAPTER 2

THE STANDARD MODEL AND ITS FUTURE

CHAPTER 3

THE CMS DETECTOR

3.1 The LHC and the CMS

3.2 Tracker of the CMS

3.3 HGCAL of the CMS

CHAPTER 4

TRIGGER STRATEGY

The analysis' signal process contains SM τ leptons in its final state. In order to exploit the leptonic decay of τ lepton, specifically with muon final state for clean signal, the B-Parking triggers are used. CMS implemented the B-Parking trigger since the year of 2018 of Run 2 for research on lepton universalities. For research of $R(K^*, D^*)$, muonic final state of B mesons are desired. B trigger requires a soft muon with modest displacement (measured using impact parameter) from the primary vertex, exploiting the b quark's long lifetime. B Parking trigger requires a muon with transverse momentum (pT) of 7-12 GeV with impact parameter (IP) 3-6. pp collisions in LHC produce extremely enormous amount of events, which could trigger the B parking trigger paths. Current CPU capacity of CMS is limited and not capable of reconstructing the entire event at such high trigger rate in HLT level. Thus, CMS scouts events, meaning it writes events that passed L1 trigger to a temporary dataset. Later, full HLT and RECO steps are implemented and served as a B-Parking dataset. The prescale factor for BPH triggers is 5-6.

4.1 Global Tags

Table 4.1: Data and MC Global tags used 2018

| | |
|-----------|-------------------------------------|
| Data 2018 | 106X_dataRun2_v29 |
| MC 2018 | 106X_upgrade2018_realistic_v11_L1v1 |

4.2 Trigger Paths

We utilize the B-Parking triggers collecting data for the year of 2018. The exact names of paths for B-Parking triggers are listed in Table 4.2. We observe that the trigger efficiency reaches a plateau after requiring a muon with pT value above that utilizes in the L1 seed.

Below is the trigger efficiency of various BPH trigger paths for different mass and lifetime points of the signal (HToSSTo4Tau) sample

Table 4.2: HLT trigger paths used in the analysis 2018.

| Data sample | Trigger |
|----------------------|--------------------|
| ParkingBPH*-Run2018A | HLT_Mu9_IP6_part* |
| ParkingBPH*-Run2018B | HLT_Mu12_IP6_part* |
| ParkingBPH*-Run2018C | HLT_Mu12_IP6_part* |
| ParkingBPH*-Run2018D | HLT_Mu12_IP6_part* |

Figure 4.1: asd

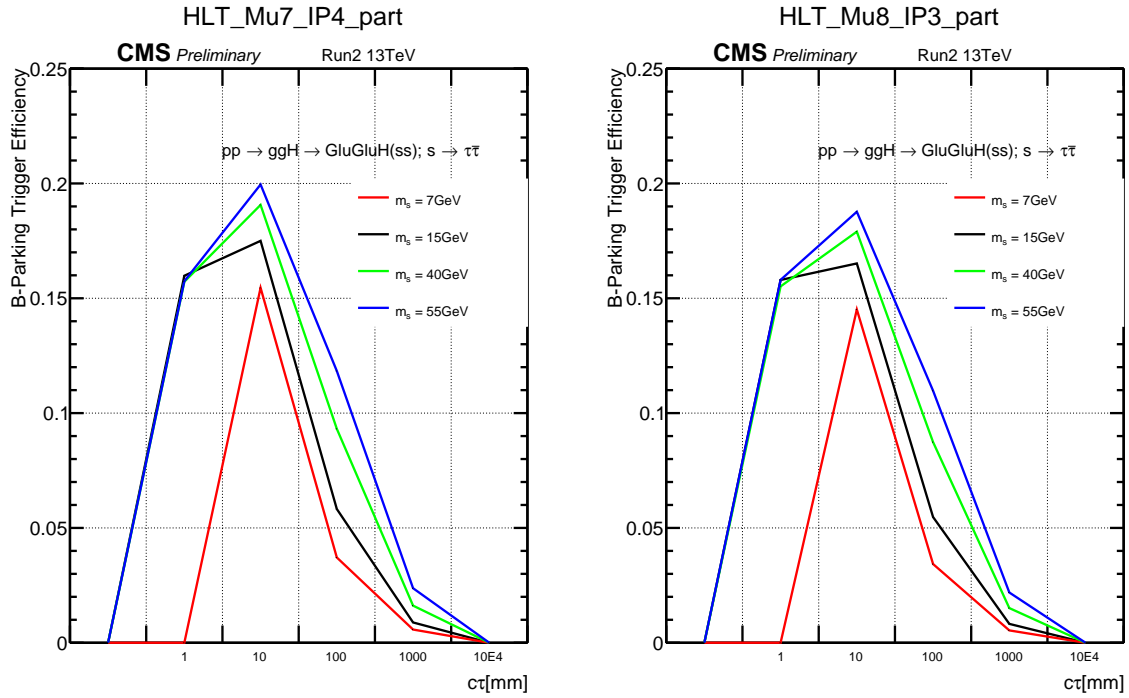
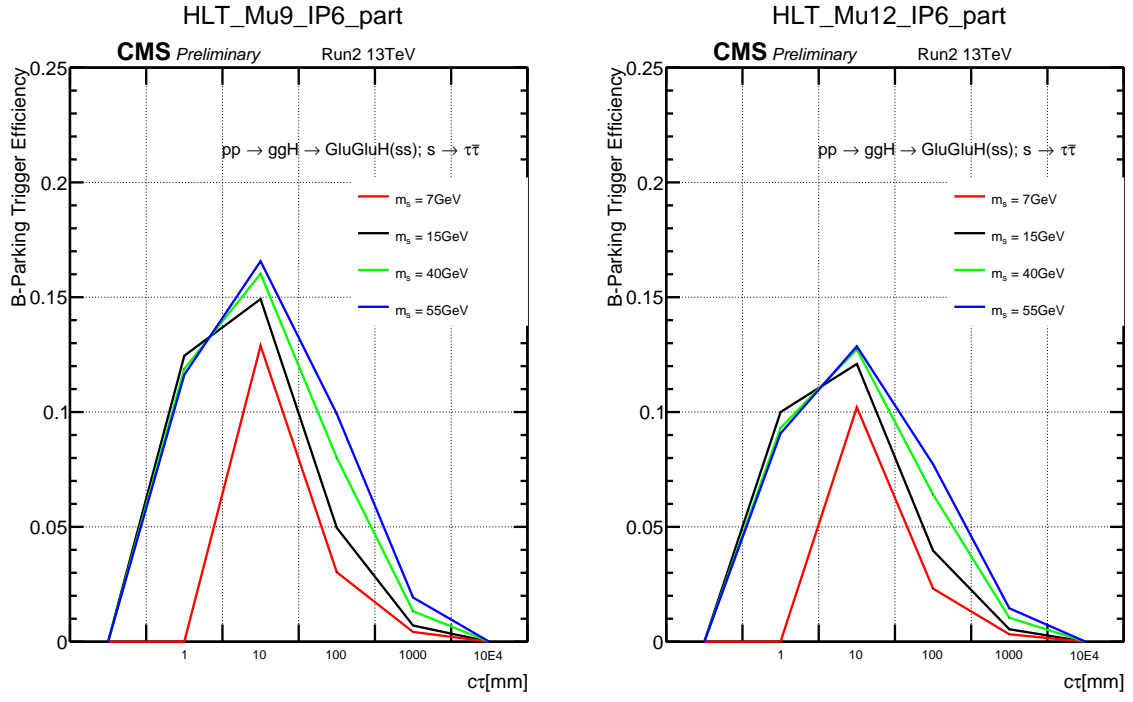


Figure 4.2: eeee



CHAPTER 5

PHYSICS OBJECT DEFINITIONS

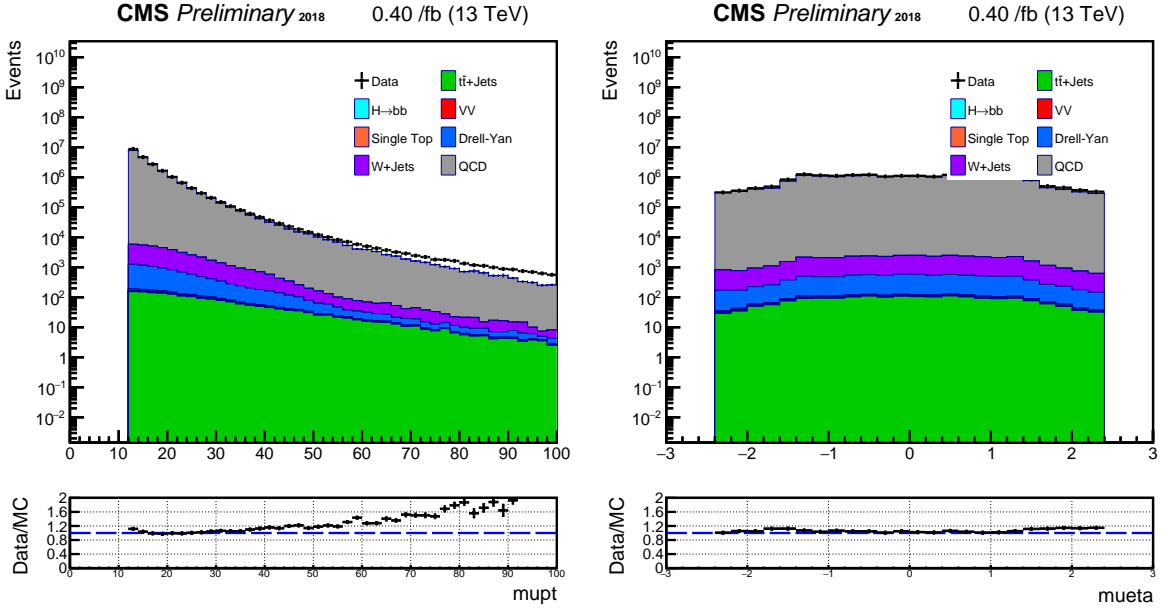
In this section, we provide the definitions of physics objects used in the analysis.

5.1 Muons

The analysis sources the SlimmedMuons collection from the MINIAOD MC datasets to produce `selectedPatMuons`. Muon objects are required to have

- $pt \geq 12$ GeV to reach BPH trigger plateau
- $|\eta| \leq 1.5$ for L1 seed $|\eta|$ cut in BPH trigger
- Pass the Loose ID criterion (`isLooseMuon`) as described in the Muon POG [50].

Figure 5.1: Data/MC of muon objects



The motivations for isolation requirements on muons are discussed in Section 7.

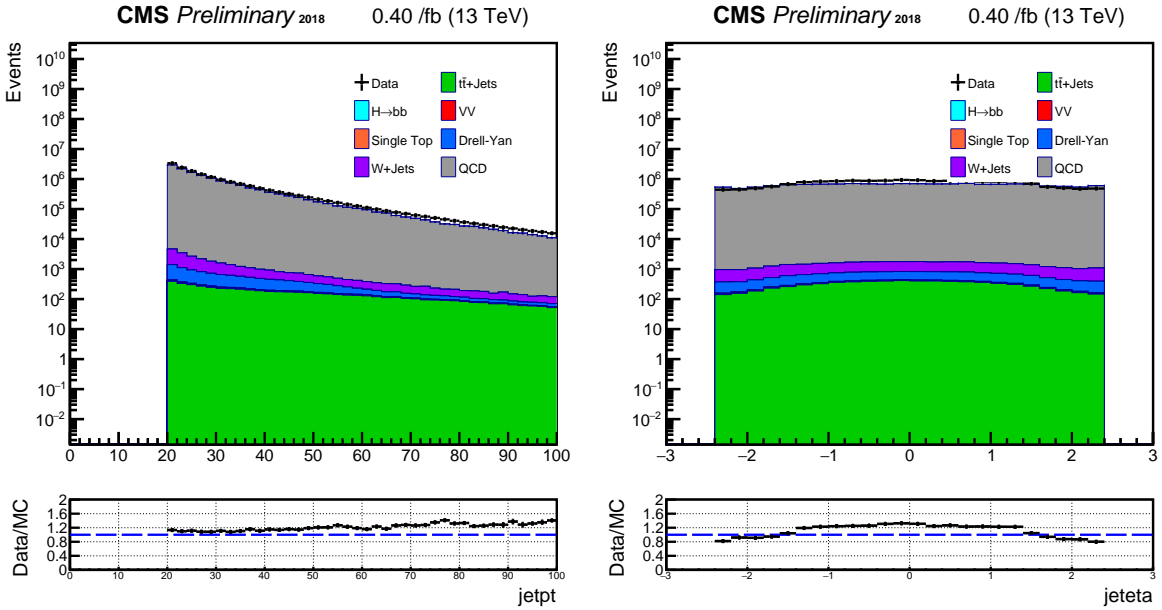
5.2 Jets

The analysis sources SlimmedJets collection from MINIAOD dataset to produce `selectedJets`. CMS reconstructs jets from calorimeter energy deposits using the anti- k_T clustering algorithm with a distance parameter of $R = 0.4$ [51]. Then, the calojets are inputted into the Particle-Flow (PF) algorithms to produce the PFJets collection. Variables in PFJets class are then slimmed to be saved into MINIAOD files. The analysis uses these SlimmedJets for the jets' b tagging scores as well. Jet objects require

- $p_t \geq 20$ GeV
- $|\eta| \leq 2.4$
- $0 \leq \text{emEnergyFraction} \leq 0.9$
- $0 \leq \text{energyFractionHadronic} \leq 0.9$
- No selected electron or muon within $\Delta R = 0.4$

The energy fraction cuts above are taken from the recommended Run2 Tight jet ID cuts for particle flow jets [52].

Figure 5.2: Data/MC of jet objects

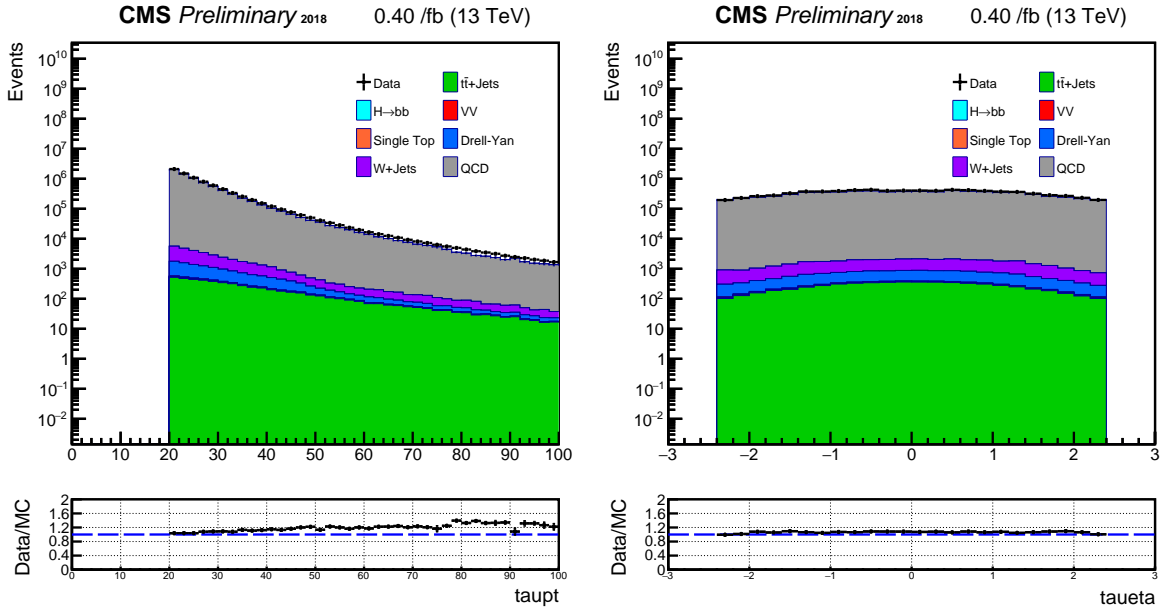


5.3 Taus

The analysis sources PAT::slimmedTaus from MINIAOD for MC and RECO::slimmedTaus for Data to produce **selectedTaus**. τ leptons have a 64% branching ratio to hadrons. The Hadron-Plus-Strip (HPS) algorithm enables the reconstruction of a τ lepton's hadronic decay. HPS uses PFJets as its starting point. τ 's hadronic decay can be reconstructed with PFJets' charged hadrons in HCAL and 2 γ s from π^0 decays in ECAL. Tau objects require

- $p_t \geq 20$ GeV
- $|\eta| \leq 2.4$

Figure 5.3: Data/MC of tau objects



5.4 Region of Interest

The complete reconstruction procedure of Regions of Interest is detailed in the following subsections. An ROI requires

- Good quality track selection
- Vertex Fitted from pair-wise tracks by V0Fitter in CMSSW
- Cluster the fitted vertices to form a Region of Interest (ROI)
- Look for tracks around $\Delta R = 0.3$ around ROI to save its isolation information

5.4.1 Tracks

The analysis sources packedPFCandidates and lostTracks from MINIAOD. Track parameters and covariance values will be propagated along the ROI production process and no value should be non-physical

- !isinf(tracks.parameter) and !isnan(tracks.parameter)
- !isinf(tracks.covariance) and !isnan(tracks.covariance)
- Number of valid hits > 3
- $pt \geq 0.35$
- Track $IPSig_{XY} \geq 2$.
- Track $IPSig_Z \geq -1$.
- Track normalized $\chi^2 \geq 10$.

5.4.2 Vertex Fitter

The analysis sources offlineBeamspt from MINIAOD for beamspot reference. Vertex fitter is KalmanVertexFitter with vertex cuts as below.

- Vertex $\chi^2 \geq 6.63$
- Transverse Decay distance significance ≥ 15 .
- $V0_{mass} \geq 13000 \text{ GeV}$
- $\cos(\theta_{XY})$ between x and p of V0 candidate ≥ 0
- $\cos(\theta_{XYZ})$ between x and p of V0 candidate ≥ -2

5.4.3 ROI formation

Fitted vertices are clustered to form a Region of Interest (ROI). These ROIs have cuts on their parameters as below.

- Radius of ROI $\geq 1 \text{ cm}$
- Annulus $\Delta R \leq 0.3$

Figure 5.4: Data/MC of ROI distribution

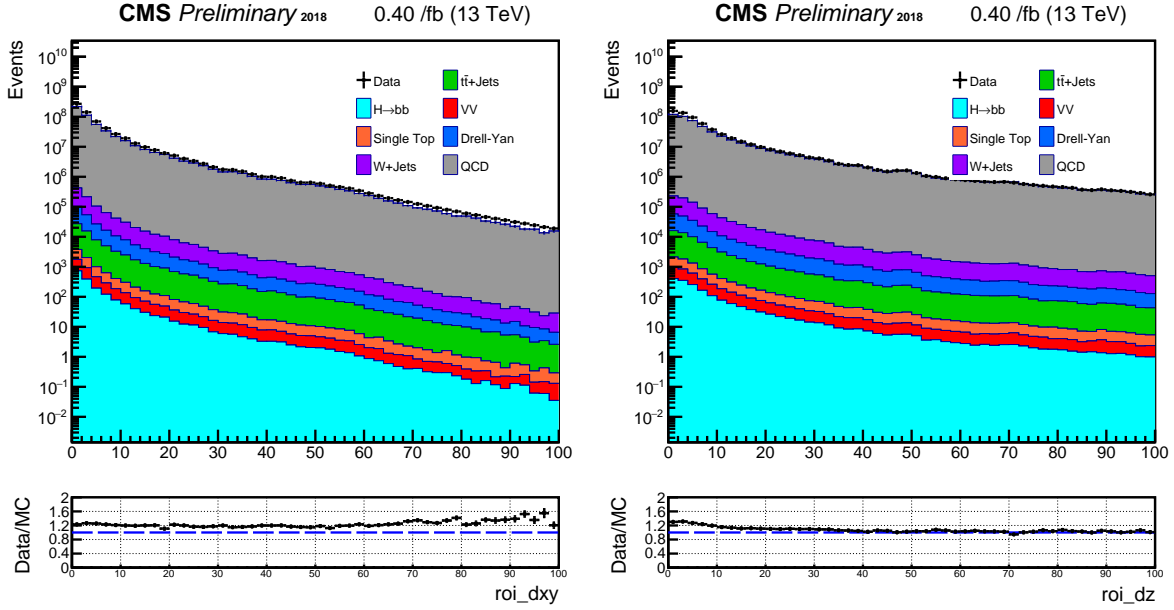
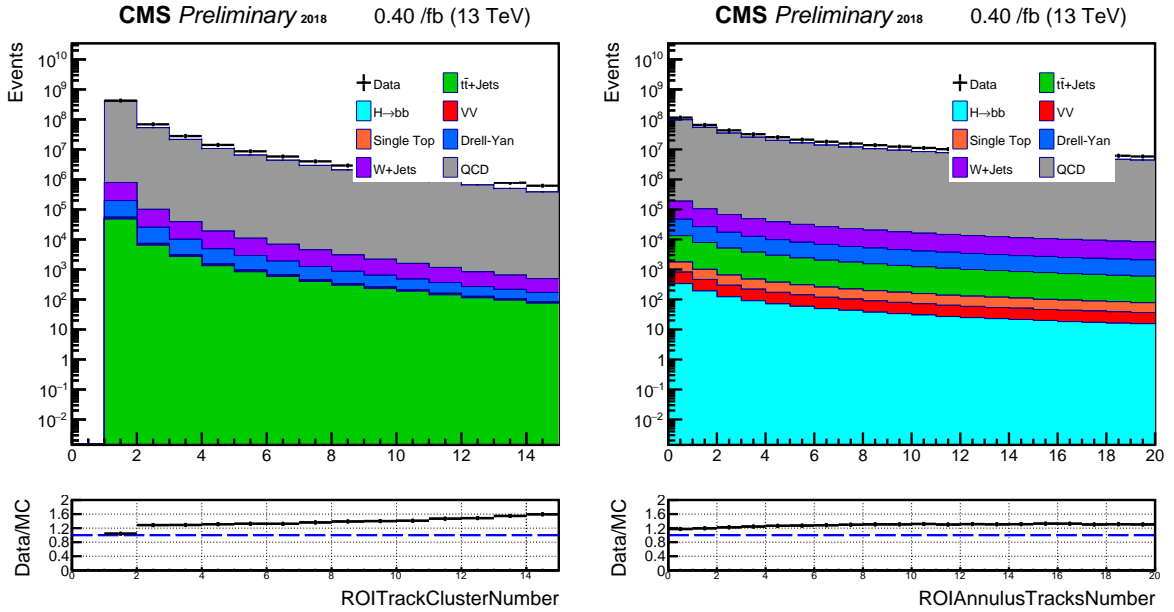


Figure 5.5: 2Data/MC of ROI distribution



5.5 Pileup Reweighting

Pileup (PU) weight values are calculated for each era of data-taking (A,B,C,D). The data's number of Pileup events is accessed for the specific HLT path with the following commands

```
brilcalc lumi --byls --normtag
/cvmfs/cms-bril.cern.ch/cms-lumi-pog/Normtags/normtag_PHYSICS.json
-i processedLumis.json --hltpath HLT_Mu9_IP6_part0_v* -o output.csv
```

Please note that we get luminosity for the specific BPH HLT path and save the output for
pileupReCalc_HLTpaths.py

```
pileupReCalc_HLTpaths.py -i output.csv
--inputLumiJSON pileup_latest.txt -o My_HLT_corrected_PileupJSON.txt
--runperiod Run2
```

where, pileup_latest.txt is from

```
/afs/cern.ch/cms/CAF/CMSCOMM/COMM_DQM/certification/Collisions18/
13TeV/PileUp/pileup_latest.txt
```

Then, we substitute the usual pileup_latest.txt to My_HLT_corrected_PileupJSON.txt to obtain
the PU distribution of data.

```
pileupCalc.py -i processedLumis.json
--inputLumiJSON My\_HLT\_corrected\_PileupJSON.txt
--calcMode true --minBiasXsec 69200 --maxPileupBin 120 --numPileupBins 120
MyDataPileupHistogram.root
```

Figure 5.6 shows the BPH1-Era A's HLT_Mu9_IP6 HLT path's Data PU distribution.

Figure 5.6: ww

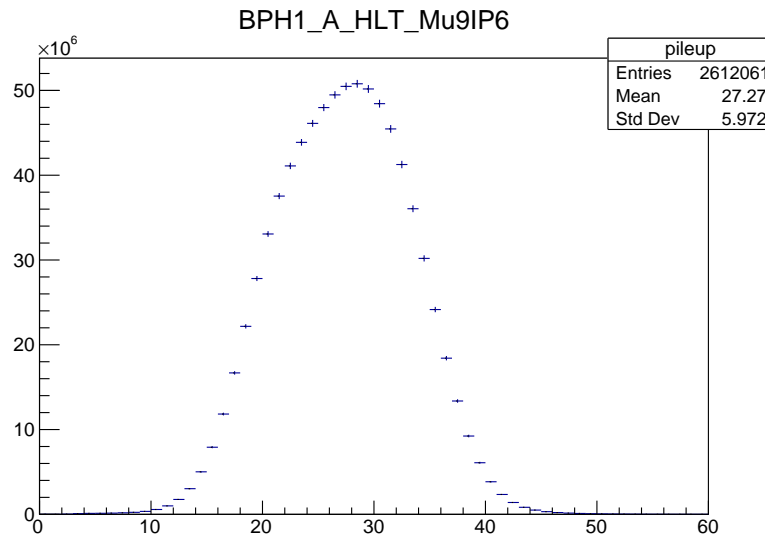


Figure 5.7 shows the BPH1-Era C's HLT_Mu12_IP6 HLT path's Data PU distribution.

Figure 5.7: dd

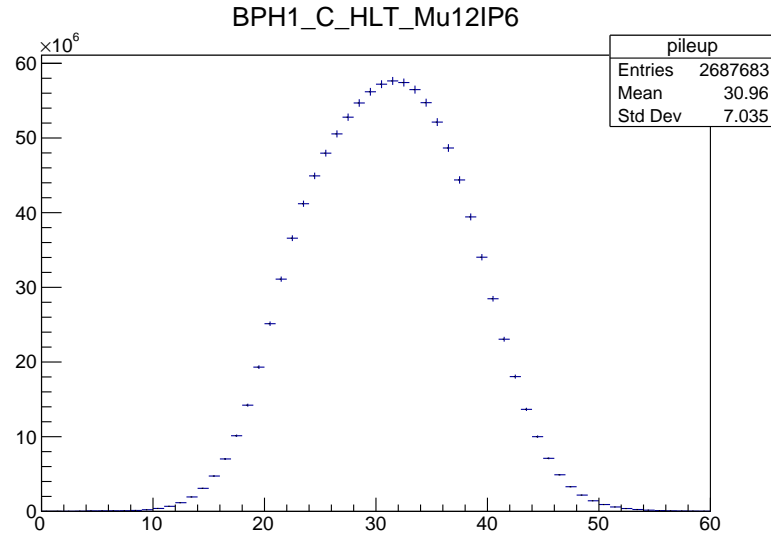


Figure 5.8: ed

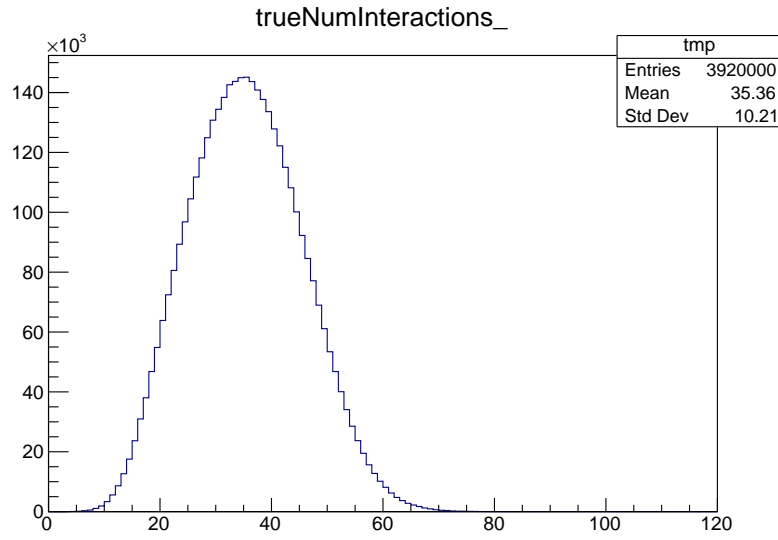
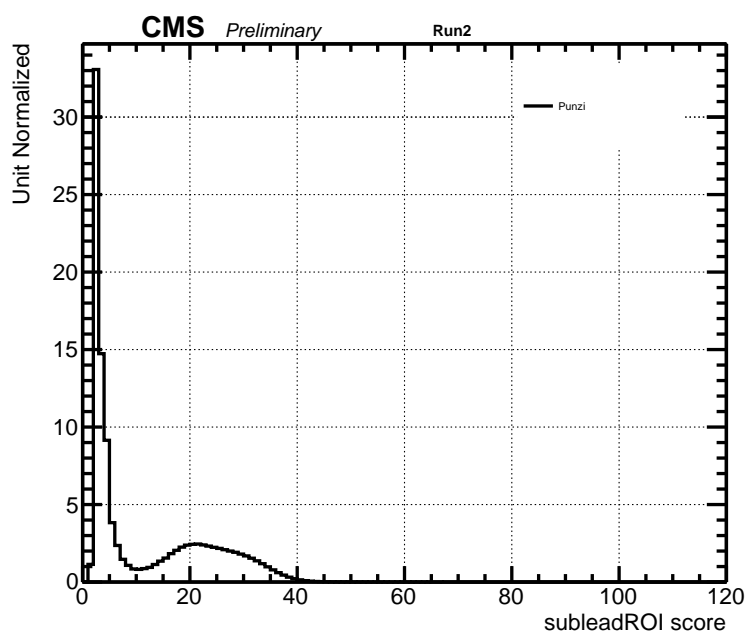


Figure 5.8 shows DYJetsToLL_M-50_TuneCP5_13TeV-madgraphMLM-pythia8 MC PU distribution.

With histograms above, one can create a PUWeight histogram distribution for an era's specific HLT path with commands below.

Figure 5.9 shows resultant such PUWeight from BPH1_A_HLT_Mu9_IP6 and DYJetsToLL_M-50_TuneCP5_13TeV-madgraphMLM-pythia8.

Figure 5.9: aaaa



CHAPTER 6

MACHINE LEARNING

ROIs are artificial regions created by the CMSSW mechanism, which are displaced vertex candidates. ROIs contain thorough information about fitted tracks, vertices, and isolation information, after following the formation procedure described in the previous section. The exhaustive variables saved in each ROI contain information, which directly and indirectly tells whether the ROI is from signal process or background process. Given the extensive amount of variables within ROIs, it's inappropriate to use ROIs' single or a few variables as our tagging variables, like in ZH analysis [48]. It is also inefficient to implement a cut-based approach due to having approximately 20-30 variables from ROI. Optimization process for all 20-30 variables would be extremely time-consuming, inefficient, and error-prone as well. Therefore, the analysis exploits machine learning (ML) for these multivariate analysis. Boosted decision trees will also face similar problem as cut-based analysis, although to a lesser extent. Deep Neural Network are the most adaptable ML algorithm for this analysis method, in which the analysis inputs extensive list of ROI variables into a neural network, and receives a final score to discriminate ROIs that arise in signal process versus background processes.

6.1 Machine Learning Software

The analysis uses Keras-Tensorflow via CMSSW. CMSSW includes Keras-Tensorflow, which enables simple and easy usage of keras-Tensorflow with simple `cmsenv` command in various CMS remote clusters. For CMSSW_10_6_12 (to have access for BPH trigger path), Keras-Tensorflow version is 2.3.1. It runs with CPUs and GPUs. CMSSW includes image container for Keras-Tensorflow and researchers can submit remote batch jobs for its ML training. The analysis tested multiple variables (such as epoch numbers, batch sizes, phi sizes, f sizes) of our DNN layers thanks to submission of remote batch jobs with CMSSW's Keras-Tensorflow image container. Subsection 6.3 below details the numerical results from such tests. The analysis also tested different combinations of signal vs background datasets for maximum discriminant power. The list of mass scale and lifetime tested for signal points are listed in subsection 6.4. Different SM physics process (and their

compositions) tested for background process are also listed in subsection 6.4 The analysis used the tensorflow protocol buffer files, which were trained with parameters and physics process as in Table 6.1.

Table 6.1: Tensorflow information

| | |
|------------|-------------------------------------|
| Epoch | 300 |
| batch size | 250 |
| Phi sizes | ((64,128,256),(32,64,128)) |
| f sizes | (256,128,32) |
| Signal | ggHSSTo4Tau-MS15GeV-c τ 100mm |
| Background | QCD_Pt120-170_MuEnriched and TTJets |

6.2 Machine Learning Input Variable

The extensive variables for input of our DNN are described and categorized in Tables 6.2, 6.3, and 6.4.

Table 6.2: ROI (trackCluster) variables by category

| | | |
|--------------|------------|--|
| TrackCluster | Position | TrackClusters.vx() - primaryVertex.X() |
| | Position | TrackClusters.vy() - primaryVertex.Y() |
| | Position | TrackClusters.vz() - primaryVertex.Z() |
| | Covariance | TrackClusters.vertexCovariance()(0,0) |
| | Covariance | TrackClusters.vertexCovariance()(0,1) |
| | Covariance | TrackClusters.vertexCovariance()(0,2) |
| | Covariance | TrackClusters.vertexCovariance()(1,0) |
| | Covariance | TrackClusters.vertexCovariance()(1,1) |
| | Covariance | TrackClusters.vertexCovariance()(1,2) |
| | Track0,1 | Track0,1.pt |
| | Track0,1 | Track0,1.eta |
| | Track0,1 | Track0,1.phi |
| | Track0,1 | Track0,1.dxy |
| | Track0,1 | Track0,1.dz |
| | Track0,1 | Track0,1.normalizedChi2 |
| | Track0,1 | Track0,1.HighPurityInt |

6.3 DNN Variable Test

For Ethan.

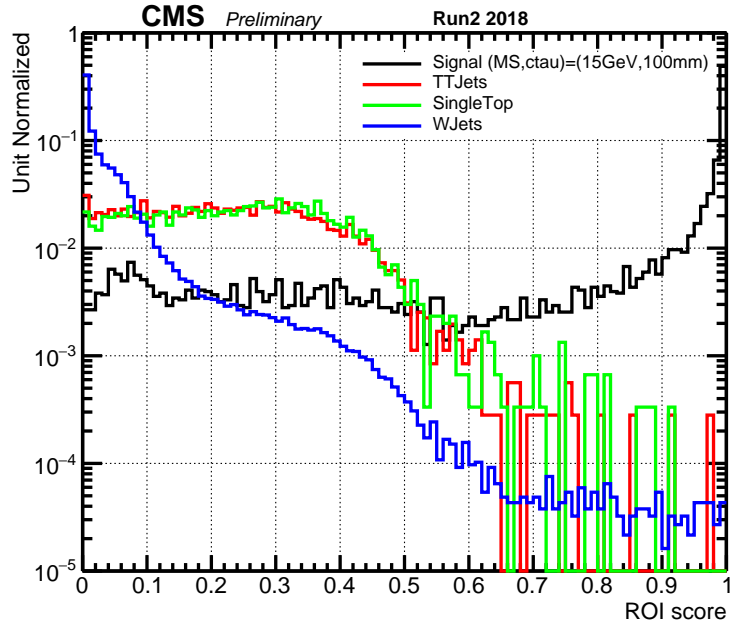
Table 6.3: ROI (Annulus) variables by category

| | | |
|---------|------------------------|--|
| Annulus | pfCandidate/LostTracks | pfCandidate/LostTracks.pt |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.eta |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.phi |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.dxy |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.dz |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.normalizedChi2 |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.HighPurityInt |
| | pfCandidate/LostTracks | pfCandidate/LostTracks.DeltaR(trackMomentum) |

Table 6.4: Event variables by category

| | | |
|-----|----------|---|
| ROI | Position | x |
| | Position | y |
| | Position | z |

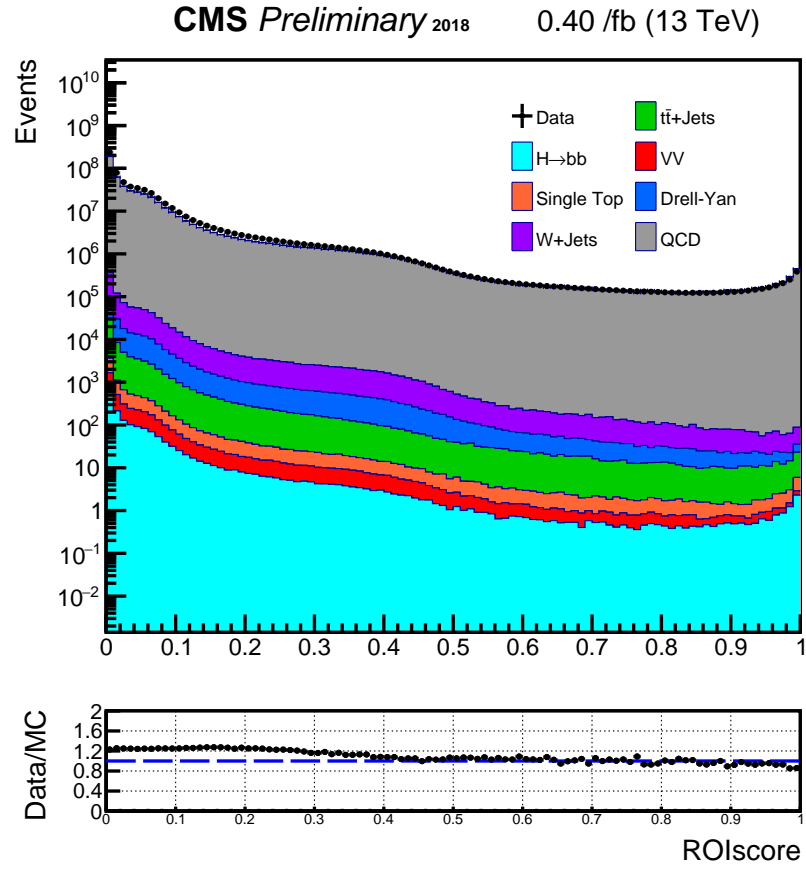
Figure 6.1: asdasdasd



6.4 Signal and Background MC Test

For Ethan.

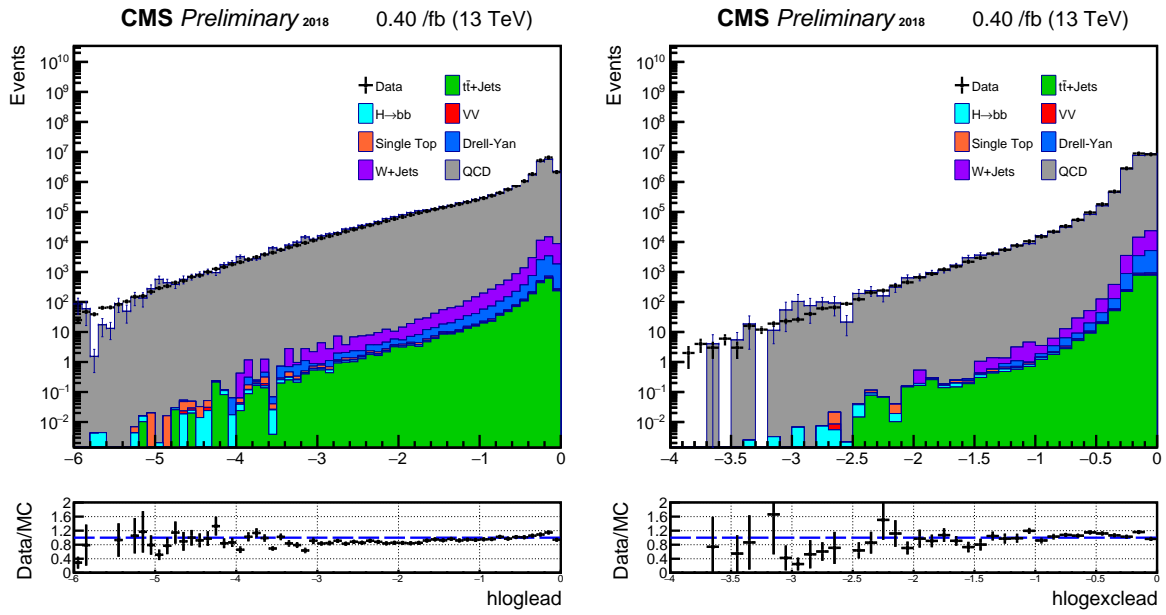
Figure 6.2: Data/MC agreement for ROI scores



6.5 SHAP values

For Ethan.

Figure 6.3: Data/MC agreement for loglead/sublead scores



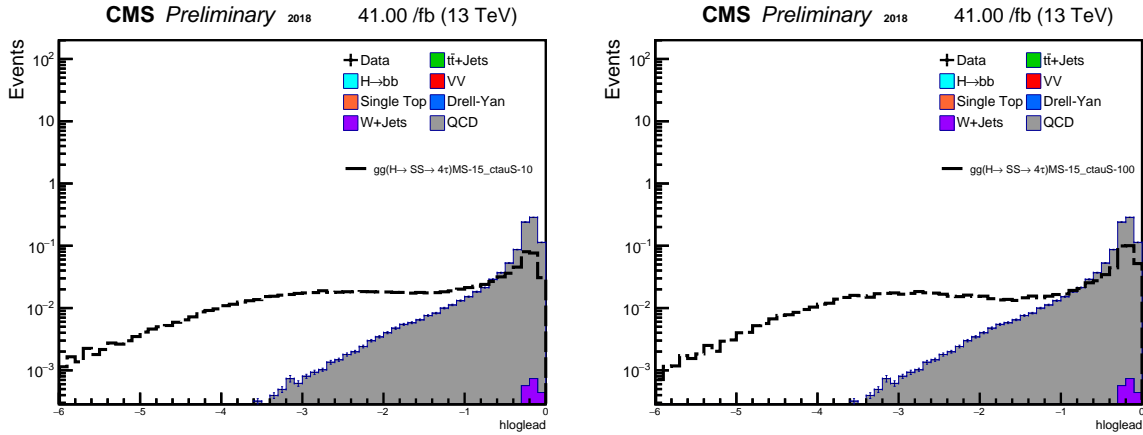
CHAPTER 7

EVENT SELECTION, SIGNAL AND CONTROL REGIONS

ROIs of all events inside the dataset listed in section ?? are scored with the tensorflow, process trained with information from section 6.2. Given that the signal process has 2 scalar decays as in Figure A.1, it's reasonable to require 2 high-scoring ROIs for our analysis. (However, the 2 selected ROIs are not simply the 2 highest-scoring ROIs of the event, due to non-negligible lifetime of τ leptons (from signal process) in the detector. The cuts on the leading ROI, and the subleading ROI are optimized based on maximizing the punzi significance formula with value σ set to discovery value of 5.

$$\sigma(N_{\text{displaced-tag}}) = \frac{S(N_{\text{displaced-tag}})}{\sqrt{B(N_{\text{displaced-tag}}) + 2.5}}. \quad (7.1)$$

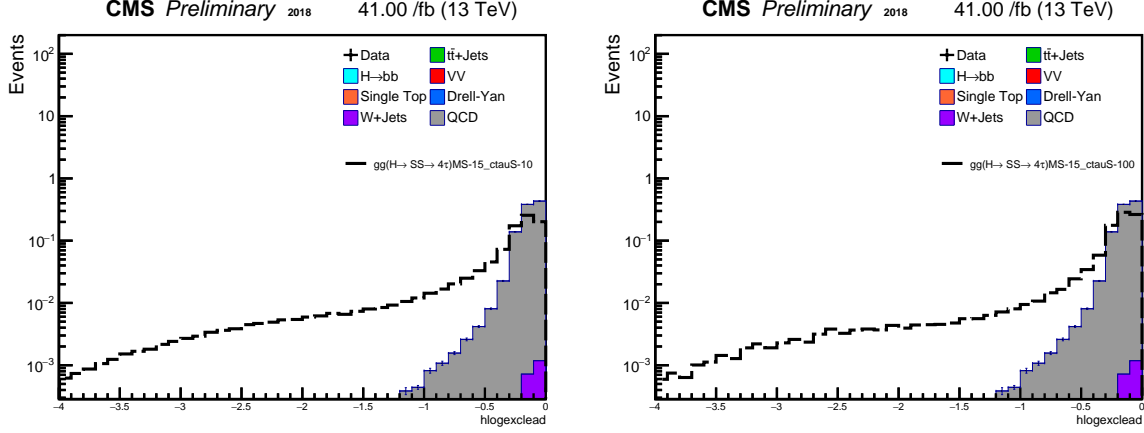
Figure 7.1: Signal versus Background for $\log_{10}(1-\text{ROIscore})$, where the ROI score is the highest ROI of the event. Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



In addition, we have other cuts applied throughout the analysis. These variables are either

- Non-relevant to ROIs (muon, jet object)
- Construction of geometric variable after selecting 2 highest scoring ROIs.
- Missing input variable in ROI training

Figure 7.2: Signal versus Background for $\log_{10}(1-\text{subROIscore})$, where the ROI score is the second highest ROI (outside of $d\Phi=0.4$ from leading ROI) of the event. Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



These are not inputs of ML training, something that can't be learned by the ML, so not discriminated with scoring of the ROIs.

They include

- $\Delta\Phi(\text{leadROI}, \text{subleadROI})$
- Number of Annulus tracks associated with $\text{ROI} < 8$
- 1 Isolated μ
- $\Delta R(\text{leadROI}, \text{Jet}) < 0.6$
- Leading μ 's transverse impact parameter to PV > 0.1

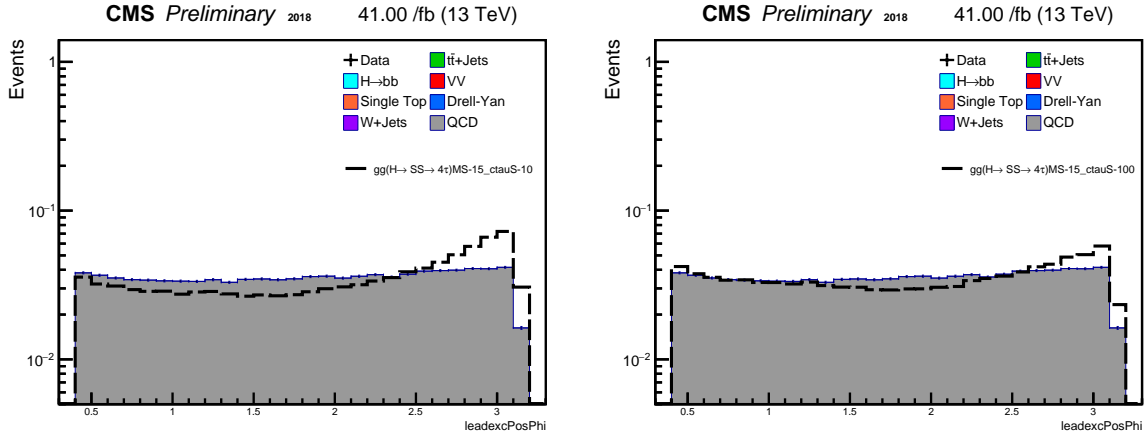
Each of the cuts for the items above is motivated and explained in following subsections.

7.1 Delta Phi(lead ROI,sublead ROI)

This analysis looks for displaced vertices in the tracker region, coming from the decays of exotic LLPs from Higgs produced in gluon fusion mode, leaving the SM Higgs boson without boost. The largest mass of the exotic LLPs is 55 GeV, ranging down to 7 GeV. Thus, exotic LLPs decayed from the SM Higgs become boosted, with their momentum vectors pointing back-to-back in the SM Higgs rest frame. Exotic LLPs with lighter mass are more boosted than heavier LLPs, since less LLP mass means more leftover energy into kinetic energy. Given that ROIs corresponding

to an exotic LLP's decay should have the highest ROI score, one should expect that the leading ROI and subleading ROI in a single event would be back-to-back. Thus, in signal events, $\Delta\Phi(\text{lead ROI, sublead ROI})$ tends to have high values, while the background processes tend to have a more uniform distribution. This analysis applies a cut above 2.2 to reduce background contribution. Optimization process for this variable is detailed in here (To be done in future?)

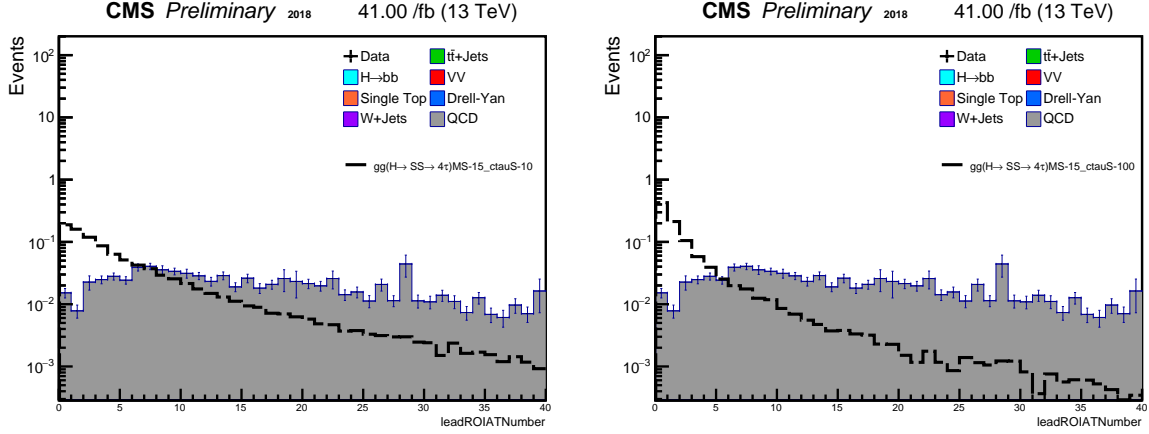
Figure 7.3: Signal versus Background for Delta Phi(leadROI, subleadROI). Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



7.2 Number of Annulus Tracks Associated with ROI

The tensorflow used for this analysis is trained with p_T, η, χ^2 , and other information of the annulus tracks (tracks that are inside ROI radius, but not fitted to vertex). Meanwhile, an ROI's total number of annulus tracks is not a direct input for ML and tensorflow can only learn such information indirectly via annulus tracks' p_T . Although having selected ROIs with high scores (>0.999), signal processes ROIs' number of annulus tracks show a quite different distribution from the background process. signal's high-scoring ROIs are mostly from the exotic LLP scalar's decay into τ leptons. Since the signal's high-score ROIs' are very well isolated, the ROI's number of annulus tracks is very low. QCD background events, which are our dominant background, have a poor isolation quality. Since QCD's high-scoring ROI's are poorly isolated due to QCD nature, these ROIs' numbers of annulus tracks are higher than the signal. More precisely, QCD's high-scoring ROI's are usually from B-mesons, which have higher track multiplicity than τ leptons.

Figure 7.4: Number of tracks in the annulus cone of the leading ROI. Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



7.3 Isolation criteria for muons

Leptonic decay of B-meson generates muons, which trigger the B-parking trigger of the analysis. Muons of B-meson decay have very poor isolation quality, just like ROIs formed around the B-meson decay. In contrast, muons of τ lepton decay have better isolation quality. In order to eliminate the dominant B-meson background from the QCD process, the analysis applies a PFISOLoose in selecting muon objects. The precise definition of PFISOLoose is defined as below.

- $(\Sigma pT(ch.hadfromPV) + \max(0, \Sigma ET(neut.had) \Sigma ET(phot) - 0.5 * \Sigma pT(ch.hadfromPU))) / P_T(\mu) < 0.25$

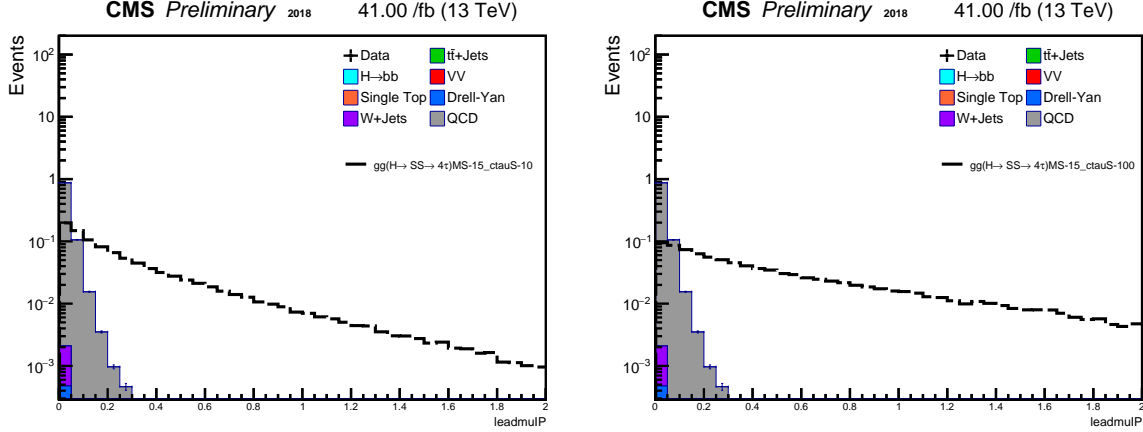
Some muons decayed from τ lepton with 7 GeV mass fail isolation cut, due to its poorer isolation quality from the boost. However, it still benefits to apply the PFISOLoose cut on muons given it removes more background events than signal events. The table below demonstrates event yield drop before and after requiring PFISOLoose cut on muons, classified by its signal and background process.

7.4 Leading muon's transverse impact parameter to PV

With the B-Parking trigger, triggering muons have significant transverse displacement (impact parameter) in both background and signal processes. However, displacement in the signal process is greater than the that of the background process. The signal process has at minimum of $c\tau = 1\text{mm}$, which is longer than B-meson lifetime. Thus, triggering muon object's transverse impact

parameter to PV is larger in signal process than background process. The analysis implements a cut on this variable.

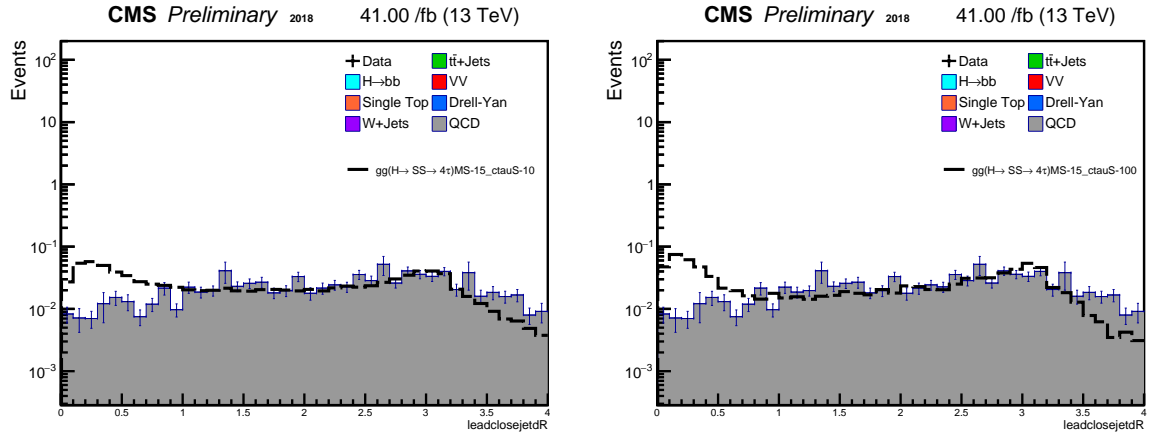
Figure 7.5: leading muon's transverse impact parameter value to the primary vertex. Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



7.5 DeltaR(ROI, jet)

τ leptons of the signal process can also decay hadronically, while only one of the τ leptons decay muonically to trigger B parking trigger. When τ leptons decay hadronically, its decay shower can get clustered in the calorimeter, and reconstructed as a jet. Given the τ lepton's on-shell mass is a fixed value, τ lepton and its hadronic decay products (to-be clustered into a jet) have a specific kinematic phase space. Thus, the $\Delta R(\text{ROI}, \text{jet})$ has a distribution with a peak at a certain value (around 0.3-0.6). Meanwhile, the QCD background has a different distribution shape. Given the hadronic nature of the process, jet multiplicity is high. Higher jet multiplicity makes the $\Delta R(\text{ROI}, \text{jet})$ value to have a rather randomized value, resulting in a flat distribution.

Figure 7.6: Delta R(Jet, leadingROI). Left plot is for MS-15_ctauS-10mm point, whereas the right plot is for MS-15_cauS-100mm point



CHAPTER 8

BACKGROUND ESTIMATION

Section 7 listed seven different discriminant variables for the analysis. Leading ROI score, subleading ROI score, and leading muon's IP value have the most discriminatory power. The analysis uses data-driven background estimation method and an ABCD method is the most preferred thanks to its simplicity. However, leading ROI score and subleading ROI score are correlated in the QCD background process. ROIs from B-mesons score higher in our tensorflow. In QCD processes, the B-mesons are likely to be pair produced. Therefore, when the leading ROI score is high due to its b-like behavior, the sublead ROI score is also high because the anti-meson is produced on the other side of the detector. Thus, leading and sublead scores can't be our ABCD discriminant variable candidates due to their correlation.

The analysis selects leading ROI and leading muon's IP value as its ABCD discriminant variables. After implementing all other cuts (sublead ROI, $\Delta\Phi(lead, sublead)$, Number of Annulus tracks associated with ROI, 1 Isolated μ , $\Delta R(leadROI, Jet)$), we tested the correlation factor between the leading ROI, and leading mu's IP values for each background process. The values are pretty minimal except for 2-3 processes where there were too few entries to derive a physical conclusion due to statistical limitations. Therefore, the analysis uses leading muIP and leading ROI score for its ABCD discriminant variables.

The results are listed below.

Figure 8.1: Cutflow histogram of MS15GeV-ct100mm point. Left plot is for region A, whereas the right plot is for region D

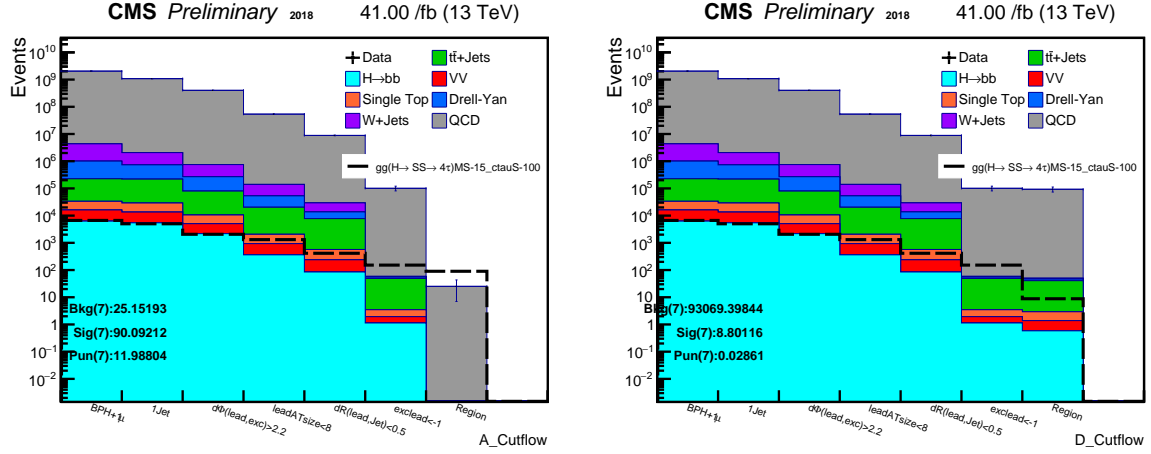
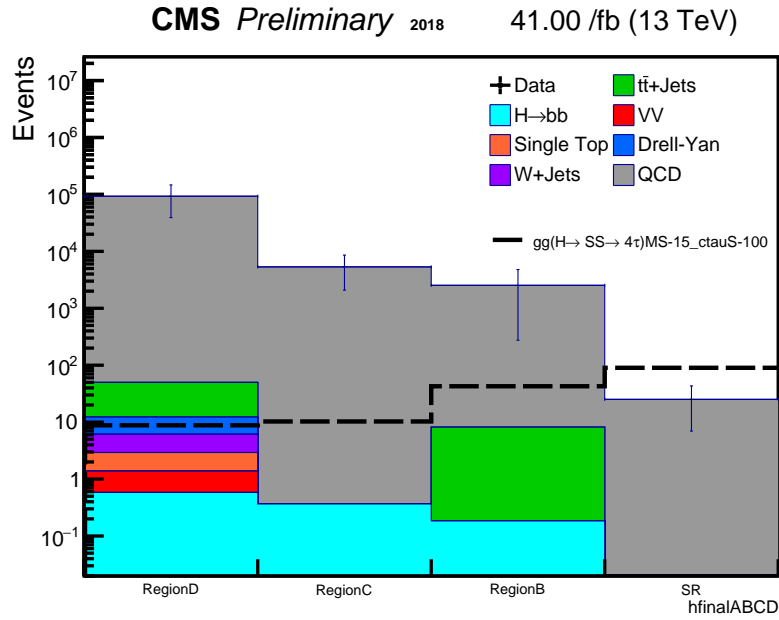


Figure 8.2: eee



CHAPTER 9

SYSTEMATIC UNCERTAINTIES

About systematic Uncertainty

CHAPTER 10

RESULTS

Results

Figure 10.1: Current Preliminary limit plots

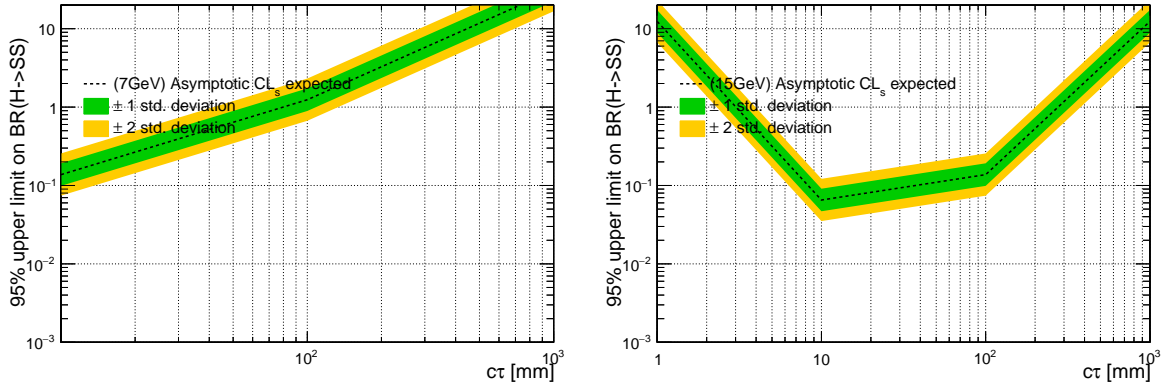
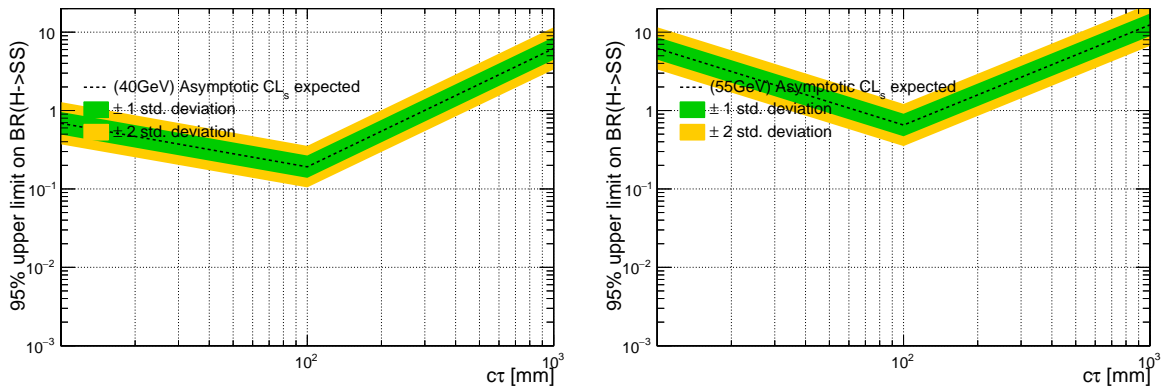


Figure 10.2: Current Preliminary limit plots



CHAPTER 11

CONCLUSIONS

This was an interesting analysis targeting tau lepton final state with tracker lifetime

APPENDIX A

DATA SAMPLES

This appendix is here merely to demonstrate how appendices may be included and formatted in your document. Look through the files `thesis.tex` and `appendix.tex` to see how these pieces work together.

The analysis uses B Parking datasets. Data was collected during the 2018 portion of Run 2 and corresponds to an integrated luminosity of 50 fb^{-1} .

Table A.1: Datasets used in the analysis

| Data sample | Integrated Luminosity (fb^{-1}) |
|---|--|
| /ParkingBPH1/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| /ParkingBPH2/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| /ParkingBPH3/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| /ParkingBPH4/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| /ParkingBPH5/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| /ParkingBPH6/Run2018A-05May2019-v1/MINIAOD | 0.866 |
| Total | 5.20 |
| /ParkingBPH1/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| /ParkingBPH2/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| /ParkingBPH3/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| /ParkingBPH4/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| /ParkingBPH5/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| /ParkingBPH6/Run2018B-05May2019-v2/MINIAOD | 1.083 |
| Total | 6.49 |
| /ParkingBPH1/Run2018C-05May2019-v1/MINIAOD | 1.079 |
| /ParkingBPH2/Run2018C-05May2019-v1/MINIAOD | 1.079 |
| /ParkingBPH3/Run2018C-05May2019-v1/MINIAOD | 1.079 |
| /ParkingBPH4/Run2018C-05May2019-v1/MINIAOD | 1.079 |
| /ParkingBPH5/Run2018C-05May2019-v1/MINIAOD | 1.079 |
| Total | 5.39 |
| /ParkingBPH1/Run2018D-05May2019promptD-v1/MINIAOD | 6.542 |
| /ParkingBPH2/Run2018D-05May2019promptD-v1/MINIAOD | 6.542 |
| /ParkingBPH3/Run2018D-05May2019promptD-v1/MINIAOD | 6.542 |
| /ParkingBPH4/Run2018D-05May2019promptD-v1/MINIAOD | 6.542 |
| /ParkingBPH5/Run2018D-05May2019promptD-v1/MINIAOD | 6.542 |
| Total | 32.7 |
| ParkingBPH Total | 50.78 |

A.1 Monte Carlo Samples

A.1.1 Signal Model and Simulation

The ggH production process (see Figure A.1) is generated at next-to-next-to-leading order (NNLO) and next-to-next-to-leading-log (NNLL) QCD and next-to-leading order (NLO) EW accuracies [53]. The Higgs boson mass is set to 125GeV for all signal samples. The cross sections, computed at NNLO+NNLL QCD and NLO EW accuracies and obtained from CERN Yellow Report 3, are 4.414 pb. The CMS detector response is modeled with GEANT4 [54].

Figure A.1: Leading Feynman diagrams for ggH production mode

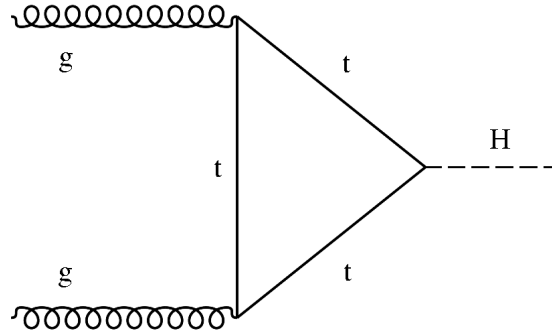


Table A.1.1 lists the signal Monte Carlo samples.

| Sample |
|---|
| /ggH_HToSSTo4Tau_MH-125_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-55_ctauS-1_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-55_ctauS-10_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-55_ctauS-100_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-55_ctauS-1000_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-40_ctauS-1_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-40_ctauS-10_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-40_ctauS-100_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-40_ctauS-1000_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-15_ctauS-1_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-15_ctauS-10_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-15_ctauS-100_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-15_ctauS-1000_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-7_ctauS-1_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-7_ctauS-10_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-7_ctauS-100_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_MS-7_ctauS-1000_TuneCP5_13TeV-powheg-pythia8/CAMPAIGN/MINIAODSIM |

A.1.2 Background Monte Carlo

All samples were processed as recommended in the PPD Run2 Analysis Guideline [55]. Tables A.2-A.5 summarizes the background Monte Carlo used in this analysis.

Figure A.2: pt of the scalar products

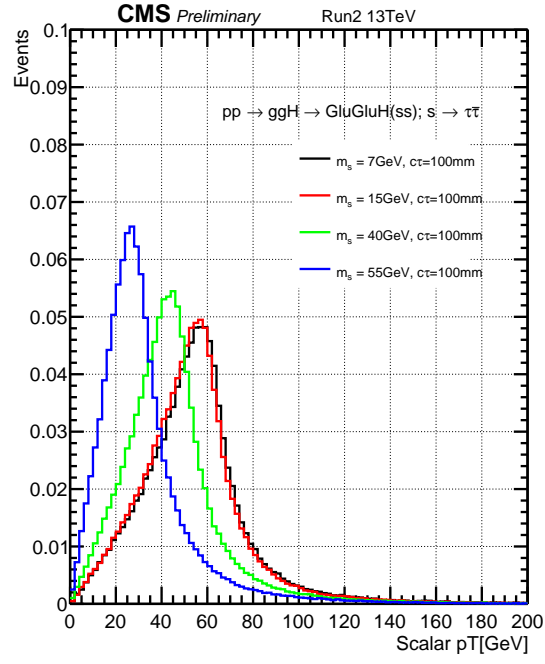
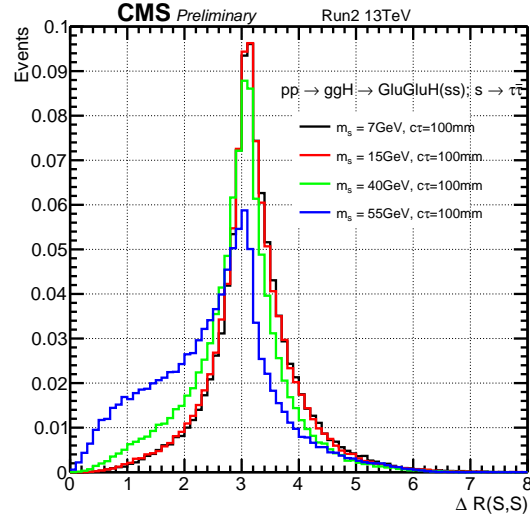


Figure A.3: DeltaR of the scalar products



a

Figure A.4: lifetime of the scalar products in the lab frame

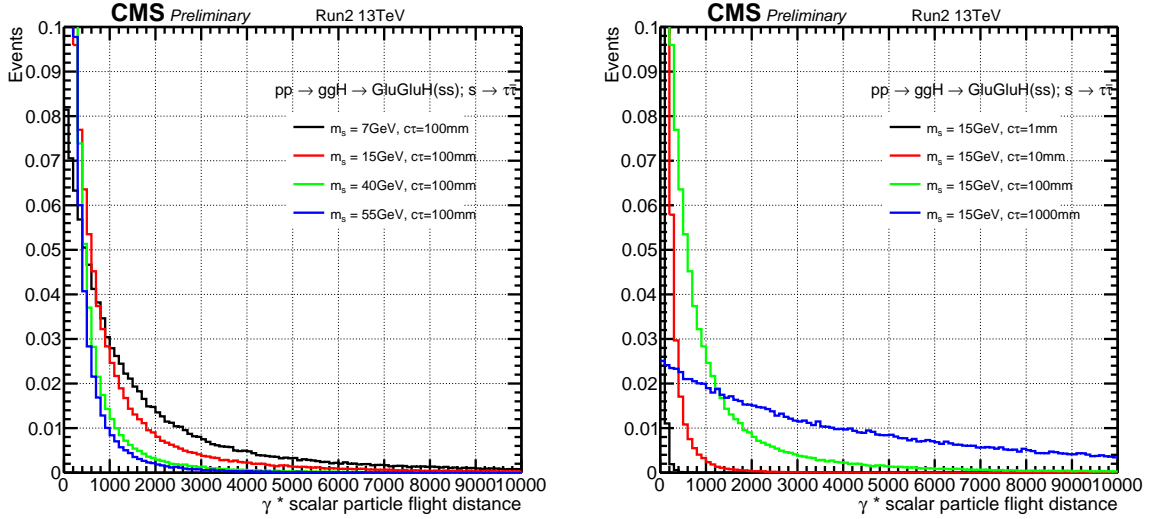


Table A.2: QCD MuEnriched Pt5 background Monte Carlo samples

| Sample |
|--|
| /QCD_Pt-15to20_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-20to30_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v4/MINIAODSIM |
| /QCD_Pt-30to50_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-50to80_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-80to120_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-120to170_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-170to300_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-300to470_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext3-v1/MINIAODSIM |
| /QCD_Pt-470to600_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-600to800_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v1/MINIAODSIM |
| /QCD_Pt-800to1000_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext3-v2/MINIAODSIM |
| /QCD_Pt-1000toInf_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v1/MINIAODSIM |

Table A.3: W,Z,H boson background Monte Carlo samples

| Sample |
|--|
| /DYJetsToLL_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v1/MINIAODSIM |
| /WJetsToLNu_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /WW_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /WZ_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v3/MINIAODSIM |
| /ZZ_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /GluGluHToBB_M125_13TeV_amcatnloFXFX_pythia8/*-v1/MINIAODSIM |

Table A.4: Top background Monte Carlo samples

| Sample |
|---|
| /TTJets_TuneCP5_13TeV-madgraphMLM-pythia8/*-v1/MINIAODSIM |
| /ST_s-channel_4f_hadronicDecays_TuneCP5_13TeV-madgraph-pythia8/*_ext1-v1/MINIAODSIM |
| /ST_t-channel_top_5f_TuneCP5_13TeV-powheg-pythia8/*-v1/MINIAODSIM |
| /ST_t-channel_antitop_5f_TuneCP5_13TeV-powheg-pythia8/*-v1/MINIAODSIM |
| /ST_tW_antitop_5f_inclusiveDecays_TuneCP5_13TeV-powheg-pythia8/*_ext1-v1/MINIAODSIM |
| /ST_tW_top_5f_inclusiveDecays_TuneCP5_13TeV-powheg-pythia8/*_ext1-v1/MINIAODSIM |

Table A.5: Monte Carlo sample summary

| Sample |
|---|
| /QCD_Pt-15to20_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-20to30_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v4/MINIAODSIM |
| /QCD_Pt-30to50_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-50to80_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-80to120_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-120to170_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-170to300_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v3/MINIAODSIM |
| /QCD_Pt-300to470_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext3-v1/MINIAODSIM |
| /QCD_Pt-470to600_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext1-v2/MINIAODSIM |
| /QCD_Pt-600to800_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v1/MINIAODSIM |
| /QCD_Pt-800to1000_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*_ext3-v2/MINIAODSIM |
| /QCD_Pt-1000toInf_MuEnrichedPt5_TuneCP5_13TeV_pythia8/*-v1/MINIAODSIM |
| /DYJetsToLL_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v1/MINIAODSIM |
| /WJetsToLNu_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /WW_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /WZ_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v3/MINIAODSIM |
| /ZZ_M-50_TuneCP5_13TeV-madgraphMLM-pythia8/*-v2/MINIAODSIM |
| /GluGluHToBB_M125_13TeV_amcatnloFXFX_pythia8/*-v1/MINIAODSIM |
| /TTJets_TuneCP5_13TeV-madgraphMLM-pythia8/*-v1/MINIAODSIM |
| /ST_s-channel_4f_hadronicDecays_TuneCP5_13TeV-madgraph-pythia8/*_ext1-v1/MINIAODSIM |
| /ST_t-channel_top_5f_TuneCP5_13TeV-powheg-pythia8/*-v1/MINIAODSIM |
| /ST_t-channel_antitop_5f_TuneCP5_13TeV-powheg-pythia8/*-v1/MINIAODSIM |
| /ST_tW_antitop_5f_inclusiveDecays_TuneCP5_13TeV-powheg-pythia8/*_ext1-v1/MINIAODSIM |
| /ST_tW_top_5f_inclusiveDecays_TuneCP5_13TeV-powheg-pythia8/*_ext1-v1/MINIAODSIM |
| /ggH_HToSSTo4Tau_MH-125_TuneCP5_13TeV-powheg-pythia8/*-v1/MINIAODSIM |

APPENDIX B

OTHERS

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BIOGRAPHICAL SKETCH

The author was born, and then the author was “educated,” at least to some degree. After finishing high school, the author completed a Bachelor of Science degree at Washington University in St.Louis. Following a decade in the work force in his discipline, the author went to FSU to pursue graduate work.