

**SEARCH FOR CHARGED-LEPTON FLAVOR VIOLATION IN
THE PRODUCTION AND DECAY OF TOP QUARKS AT
 $\sqrt{s} = 13$ TEV WITH THE CMS DETECTOR**

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**A DISSERTATION SUBMITTED IN PARTIAL FULFILLMENT OF THE
REQUIREMENTS FOR THE DEGREE OF DOCTOR OF PHILOSOPHY**

**DEPARTMENT OF PHYSICS
NORTHEASTERN UNIVERSITY**

2023

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Abstract

A search for charged-lepton flavor violation has been performed in the top quark sector through both top quark production and decay signal processes. The data were collected by the CMS experiment from proton-proton collisions at a center-of-mass energy of 13 TeV and correspond to an integrated luminosity of 138 fb^{-1} . The selected events are required to contain one opposite-sign electron-muon pair, a third charged lepton (electron or muon), at least one jet, and at most one jet associated with a bottom quark. The analysis utilizes boosted decision trees to separate background processes from a possible signal, exploiting differences in the kinematics of the final state particles. The data are found to be consistent with the standard model expectation. Upper limits at 95% confidence level are placed on the branching fractions involving up (charm) quarks, $t \rightarrow e\mu u$ ($t \rightarrow e\mu c$), of 0.032×10^{-6} (0.498×10^{-6}), 0.022×10^{-6} (0.369×10^{-6}), and 0.012×10^{-6} (0.216×10^{-6}) for tensor, vector, and scalar interactions, respectively.

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List of Acronyms

AR	application region
ATLAS	A Toroidal LHC Apparatus
BDT	Boosted Decision Tree
CLFV	Charged-Lepton Flavor Violation
CMS	Compact Muon Solenoid
EFT	Effective Field Theory
HLT	High Level Trigger
HL-LHC	High Luminosity-LHC
LHC	Large Hadron Collider
LO	leading order
L1	Level-1
MC	Monte-Carlo
ME	matrix-element
MET	missing transverse momentum
MR	measurement region
MVA	multivariate analysis
NLO	next-to-leading order
NN	neural network
NNLO	next-to-next-to-leading order
OSDF	opposite-sign and different-flavor
OSSF	opposite-sign and same-flavor
PD	Primary Dataset
PDF	Parton Distribution Function
PF	particle flow

PU pile-up

PV primary vertex

QCD quantum chromodynamics

SM Standard Model

SR signal region

SSSF same-sign and same-flavor

SV secondary vertex

VR validation region

WC Wilson Coefficient

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[1]

CHAPTER 12

Datasets, Simulated Samples and Triggers

This analysis is based on data collected by the Compact Muon Solenoid (CMS) experiment in 2016-2018 from pp collisions at a center-of-mass energy of 13 TeV corresponding to an integrated luminosity of 138 fb^{-1} . There were approximately 30 simultaneous pp collisions occurring per 25 ns. Based on online selection criteria, fully reconstructed collision data that contain high-level physics objects are divided into “Primary Datasets (PDs)”. The PDs that make use of lepton information for selection include “DoubleEG”, “DoubleMu”, “MuonEG”, “SingleElectron”, and “SingleMuon” for 2016 and 2017 data taking era. In 2018, “SingleElectron”, “DoubleEG” are replaced by “EGamma”. The names of these PDs reflect the selection criteria. In addition to these PDs, MC samples are also generated to model both signal and background processes, which are described in section 12.1 and section 12.2, respectively. To account for the different data taking conditions across the years, all MC samples are generated separately for each year. HLT triggers are used to select events offline, which is described in section 12.3.

12.1 Signal Samples

In this analysis, New Physics is described by Dimension-6 Effective Field Theory (EFT) operators,

$$\mathcal{L} = \mathcal{L}_{\text{SM}}^{(4)} + \frac{1}{\Lambda^2} \sum_a C_a^{(6)} O_a^{(6)} + O\left(\frac{1}{\Lambda^4}\right). \quad (12.1)$$

Among many of the Dimension-6 operators in Warsaw basis [2], a total of 6 operators are considered, which are summarized in Table 12.1. To reduce the number of free parameters, the permutations of fermion flavors are combined. Taking $e\mu\tau\nu$ vertex as an example, the Wilson Coefficients (WCs) are parameterized in the following way:

$$C_{lq} = C_{lq}^{(1)1213} + C_{lq}^{(1)2113} + C_{lq}^{(1)1231} + C_{lq}^{(1)1213}, \quad (12.2)$$

$$C_{lu} = C_{lu}^{1213} + C_{lu}^{2113} + C_{lu}^{1231} + C_{lu}^{1213}, \quad (12.3)$$

$$C_{eq} = C_{eq}^{1213} + C_{eq}^{2113} + C_{eq}^{1231} + C_{eq}^{1213}, \quad (12.4)$$

$$C_{eu} = C_{eu}^{1213} + C_{eu}^{2113} + C_{eu}^{1231} + C_{eu}^{1213}, \quad (12.5)$$

Table 12.1: Summary of relevant dimension-6 operators considered in this analysis. Here, ε is the two dimensional Levi-Civita symbol, γ^μ the gamma matrix, and $\sigma^{\mu\nu} = \frac{i}{2}[\gamma^\mu, \gamma^\nu]$. The l and q denote left-handed doublets, whereas u and e denote right-handed singlets. The indices i and j are lepton flavor indices that run from 1 to 2 with $i \neq j$; m and n are quark flavor indices with the condition that one of them is 3 and the other one is 1 or 2.

Lorentz Structure	Operator
vector	$O_{lq}^{(1)ijmn} = (\bar{l}_i \gamma^\mu l_j)(\bar{q}_m \gamma_\mu q_n)$
	$O_{lu}^{ijmn} = (\bar{l}_i \gamma^\mu l_j)(\bar{u}_m \gamma_\mu u_n)$
	$O_{eq}^{ijmn} = (\bar{e}_i \gamma^\mu e_j)(\bar{q}_m \gamma_\mu q_n)$
	$O_{eu}^{ijmn} = (\bar{e}_i \gamma^\mu e_j)(\bar{u}_m \gamma_\mu u_n)$
scalar	$O_{lequ}^{(1)ijmn} = (\bar{l}_i e_j) \varepsilon (\bar{q}_m u_n)$
tensor	$O_{lequ}^{(3)ijmn} = (\bar{l}_i \sigma^{\mu\nu} e_j) \varepsilon (\bar{q}_m \sigma_{\mu\nu} u_n)$

$$C_{lequ}^{(1)} = C_{lequ}^{(1)1213} + C_{lequ}^{(1)2113} + C_{lequ}^{(1)1231} + C_{lequ}^{(1)1213}, \quad (12.6)$$

$$C_{lequ}^{(3)} = C_{lequ}^{(3)1213} + C_{lequ}^{(3)2113} + C_{lequ}^{(3)1231} + C_{lequ}^{(3)1213}. \quad (12.7)$$

Additionally, all vector-like operators are combined,

$$O_{e\mu tu}^{\text{vector}} = O_{lq} + O_{lu} + O_{eq} + O_{eu}, \quad (12.8)$$

$$O_{e\mu tu}^{\text{scalar}} = O_{lequ}^{(1)} + \text{h.c.}, \quad (12.9)$$

$$O_{e\mu tu}^{\text{tensor}} = O_{lequ}^{(3)} + \text{h.c.}, \quad (12.10)$$

which results in 6 independent **WCs**: $C_{e\mu tu}^{\text{vector}}$, $C_{e\mu tu}^{\text{scalar}}$, $C_{e\mu tu}^{\text{tensor}}$, $C_{e\mu tc}^{\text{vector}}$, $C_{e\mu tc}^{\text{scalar}}$, $C_{e\mu tc}^{\text{tensor}}$.

To generate signal **MC** samples, the effective Lagrangian described above is implemented using the SmeftFR v2 [3] model, and saved in the “UFO” format [4]. Additionally, all the **WCs** are set to 1 with $\Lambda = 1$ TeV in the UFO, which then interfaces with the FEYNRULES [5] package to calculate Feynman diagrams. The output of the FEYNRULES is used in **ME** event generator MADGRAPH5_AMC@NLO v2.4.2 [6] to generate events at leading order (**LO**).

In general, the calculations done by the **ME** event generators are model-agnostic assuming the same **EFT** configurations. In other words, models like SmeftFR or SMEFTsim [7] are expected to give the same or very similar results in terms of cross sections and four-momenta of final-

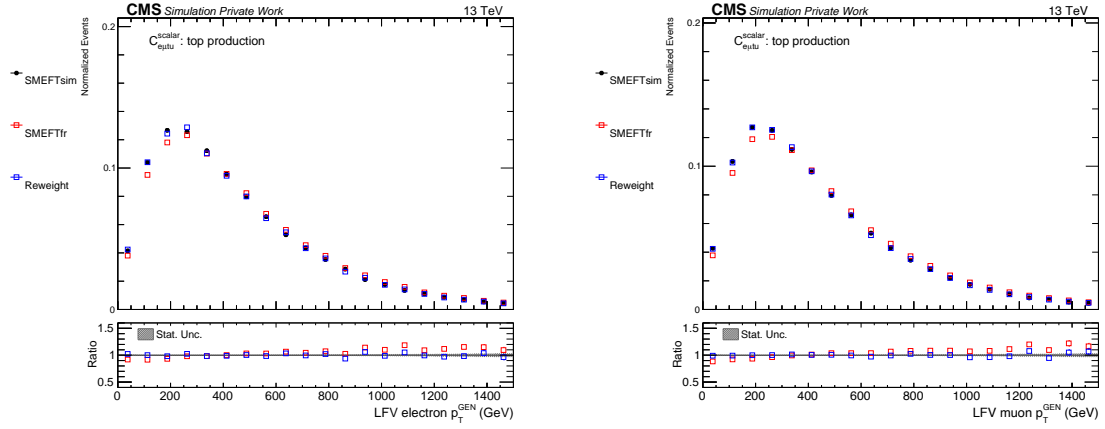


Figure 12.1: Comparison of kinematic distributions at **ME**-level produced by different models: LFV electron p_T (left), LFV muon p_T (right). The “SmeftFR” samples (shown in red curve) and “SMEFTsim” samples (shown in black curve) are statistically independent of each other. The “Reweight” (shown in blue curve) are produced by applying weights calculated by Equation 12.11 to “SmeftFR” samples.

state particles. Nevertheless, visible differences in kinematics have been observed and shown in Figure 12.1. Furthermore, the cross sections predicted by SmeftFR v2 also yield more than 20% difference relative to SMEFTsim due to a bug that was later fixed in SmeftFR v3. In the light of these differences, the **CMS** and A Toroidal LHC Apparatus (**ATLAS**) Collaborations agreed to adopt the SMEFTsim model as the common standard. To quantify the impact of the choice of models on kinematics, the following ratio is calculated for each event i ,

$$R_{\text{reweight}}^i = \frac{\omega_{\text{SMEFTsim}}^i}{\omega_{\text{SmeftFR}}^i}, \quad (12.11)$$

where ω_X^i is the per-event **ME** weight calculated by MADGRAPH5_AMC@NLO using model X . Since SMEFTsim was not used by **CMS** at the time when the signal samples were generated, R_{reweight} are used to “reweight” the original samples generated using SmeftFR.

Due to the significant differences in kinematic distributions between top decay and production signals, **MC** samples are generated separately for these processes. The cross sections for top production signals are taken directly from MADGRAPH5_AMC@NLO with SMEFTsim UFO as input. The event generation for top decay signals at the **ME**-level take two steps: (i) production of the SM $t\bar{t}$, and (ii) **CLFV** decay of one of the top quarks. Therefore, the $t\bar{t}$ cross section at next-to-next-to-leading order (**NNLO**) precision [8] is used to normalize the top decay signals,

$$\sigma_{\text{CLFV}}^{\text{Top Decay}} = 2 \times \sigma_{\text{tt}}^{\text{NNLO}} \times \mathcal{B}(t \rightarrow e\mu q), \quad (12.12)$$

where $q=\{u,c\}$, and $\mathcal{B}(t \rightarrow e\mu q)$ [9] can be expressed as,

$$\mathcal{B}(t \rightarrow e\mu q) = \frac{\frac{|C_{e\mu tq}^{\text{vector}}|^2}{\Lambda^4} \frac{m_t^5}{384\pi^3\Gamma_t^{\text{SM}}}}{\frac{|C_{e\mu tq}^{\text{scalar}}|^2}{\Lambda^4} \frac{m_t^5}{3072\pi^3\Gamma_t^{\text{SM}}} + \frac{|C_{e\mu tq}^{\text{tensor}}|^2}{\Lambda^4} \frac{m_t^5}{64\pi^3\Gamma_t^{\text{SM}}}} \quad (12.13)$$

where m_t and Γ_t^{SM} are taken to be 172.5 GeV and 1.33 GeV in this analysis, respectively. The choice of u or c quark in final states does not affect the cross sections of the top decay signals. The cross sections for all signal **MC** samples are summarized in Table 12.2.

Table 12.2: Theoretical cross sections for top production and decay for each **CLFV** coupling, calculated at $C/\Lambda^2 = 1 \text{ TeV}^{-2}$. Uncertainties related to PDF and QCD scale in **ME** calculation are given ($\sigma_{\text{scale}}^{+\text{scale}} \pm \text{PDF}$).

Lorentz Structure	Samples	XS (fb)
vector	top production via u quark	$460_{-64}^{+81} \pm 6$
	top production via c quark	$33_{-4}^{+5} \pm 6$
	top decay via u/c quark	$32_{-1.1}^{+0.8} \pm 1.3$
scalar	top production via u quark	$97_{-14}^{+18} \pm 1$
	top production via c quark	$6.3_{-0.8}^{+0.9} \pm 1.4$
	top decay via u/c quark	$4.0_{-0.1}^{+0.1} \pm 0.2$
tensor	top production via u quark	$2143_{-293}^{+368} \pm 31$
	top production via c quark	$164_{-18}^{+22} \pm 27$
	top decay via u/c quark	$187_{-6}^{+5} \pm 8$

Steps other than the **ME** calculation concerning signal **MC** generation follow the **CMS** standard, which is described in the following section.

12.2 Background Samples

Besides tZq , tHq , tHW , and tWZ processes, the next-to-leading order (**NLO**) Parton Distribution Function (**PDF**) set from NNPDF3.0 [10] is used in 2016 to generate background **MC** samples. The **NNLO PDF** set from NNPDF3.1 [11] is used for tZq while the **LO PDF** set from NNPDF3.0

Table 12.3: Summary of the configurations of the MC samples. DYM50 (DYM10to50) denote DY sample with a dilepton invariant mass greater than 50 GeV (between 10 and 50 GeV). V includes W and Z bosons. The cross sections for samples without a citation are taken directly from their event generators.

Category	Process	Event Generator	Perturbative QCD	Tune	XS precision
<i>prompt</i> background	WZ	MADGRAPH	NLO	CUETP8M1(CP5)	NLO [16]
	ZZ	POWHEG	NLO	CUETP8M1(CP5)	NLO [16]
	VVV	MADGRAPH	NLO	CUETP8M1(CP5)	NLO
	$t\bar{t}W$, $t\bar{t}Z$	MADGRAPH	NLO	CUETP8M1(CP5)	NLO [17, 18]
	$t\bar{t}H$	POWHEG	NLO	CUETP8M1(CP5)	NLO [18]
	tZq	MADGRAPH	NLO	CP5	NLO
	tHq , tHW , tWZ	MADGRAPH	LO	CUETP8M1(CP5)	LO
<i>nonprompt</i> background	$t\bar{t}$	POWHEG	NLO	CUETP8M1(CP5)	NNLO [8]
	DYM50	MADGRAPH	NLO	CUETP8M1(CP5)	NNLO [19]
	DYM10to50	MADGRAPH	LO	CUETP8M1(CP5)	NLO [19]

is used for tHq , tHW , and tWZ in 2016. In 2017 and 2018, the NNLO PDF set from NNPDF3.1 is used to generate all the samples.

The default choice of ME event generator is MADGRAPH5_AMC@NLO v2.4.2 (v2.2.2 for 2016), which is used to generate all but ZZ, $t\bar{t}H$, and $t\bar{t}$ samples. These three samples are generated with POWHEG v2 [12] instead. Samples with small contributions (tHq , tWZ , tHW , and low mass DY) are generated at LO while other samples are generated at NLO. Whenever possible and relevant, theoretical cross sections from high-order quantum chromodynamics (QCD) calculations are used. The references of these calculations are included in 12.3.

The PYTHIA v8.2 [13] is used to model parton shower and hadronization. The CUETP8M1 [14] is used in 2016 for underlying event tuning while the CP5 [15] is used in 2017 and 2018. The configurations of the MC samples are summarized in Table 12.3. The background processes are divided into two categories: (i) processes with three or more *prompt* leptons in the final states are classified as “*prompt* background”, and (ii) other processes are classified as “*nonprompt* background”. The *nonprompt* backgrounds in this analysis are modeled with a data-driven technique, which is discussed in chapter 15. The MC samples listed in “*nonprompt*” category in Table 12.3 are therefore only used for the purpose of validations.

12.3 Triggers

The target final states of this analysis contain three prompt leptons, which make lepton triggers the most optical choice to select events. To achieve the maximum acceptance, a combination of single-lepton, di-lepton and tri-lepton triggers are used. These triggers are summarized in [Appendix A](#). Events in simulated samples are required to fire at least one of the triggers listed in [Table A.1-A.3](#). Since multiple [PDs](#) are used to record data events and the orthogonality of these [PDs](#) are not guaranteed by the online selection criteria, the following trigger logic is implemented to remove the overlap between different [PDs](#):

- Events in SingleMuon datasets are required to fire at least one of the triggers listed under “SingleMuon”.
- Events in DoubleMuon datasets are required to fire at least one of the triggers listed under “DoubleMu”. Events are removed if they also fire at least one of the triggers listed under “SingleMuon”.
- Events in MuonEG’ datasets are required to fire at least one of the triggers listed under “MuonEG”. Events are removed if they also fire at least one of the triggers listed under “SingleMuon” or “DoubleMu”.
- Events in Single Electron datasets are required to fire at least one of the triggers listed under “SingleElectron”. Events are removed if they also fire at least one of the triggers listed under “SingleMuon” or “DoubleMu” or “MuonEG”.
- Events in DoubleEG datasets are required to fire at least one of the triggers listed under “DoubleEG”. Events are removed if they also fire at least one of the triggers listed under “SingleMuon” or “DoubleMu” or “MuonEG” or “SingleElectron”.
- Events in EGamma datasets are required to fire at least one of the triggers listed under “EGamma”. Events are removed if they also fire at least one of the triggers listed under “SingleMuon” or “DoubleMu” or “MuonEG”.

CHAPTER 13

Object Selection

Objects described in [chapter 8](#), referred to as “candidates”, are further selected with more stringent requirements with the goal of suppressing the contributions from background processes while maintaining a high signal acceptance. In particular, prompt electron and muon candidates are identified through a custom-trained Boosted Decision Tree ([BDT](#)) classifier, which is discussed in [section 13.1](#). Two jet identification algorithms are deployed to select jet candidates originating from hard collisions, which is discussed in [section 13.2](#). Furthermore, jet candidates that originate from b quarks are identified with a neural network ([NN](#)) based algorithm, which is discussed in [section 13.3](#).

13.1 Lepton Selection

The target final states of this analysis feature exactly three leptons that originate either from decays of electroweak bosons or from the [CLFV](#) interaction, which in this analysis is a contact interaction that involves four fermions. These leptons, referred to as *prompt* leptons, typically appear to be isolated and not far away from the primary vertex ([PV](#)). In contrast, *nonprompt* leptons are leptons that originate from decays of hadrons, or from photon conversions, or misidentified leptons. They often travel a noticeable distance away from the [PV](#) and appear to be less isolated due to nearby activities. Due to the high multiplicity of leptons in our selection, backgrounds with at least one *nonprompt* lepton outnumber any other [SM](#) processes that produce three or more *prompt* leptons. It is therefore crucial to exploit the differences between *nonprompt* and *prompt* leptons and bring the *nonprompt* background under control.

13.1.1 TOP LeptonMVA

The TOP LEPTONMVA is an offline lepton identification algorithm that is originally developed for tZq analyses [[20](#), [21](#)]. It is based on Gradient [BDT](#) implemented using the TMVA package [[22](#)]. A total of 13 features are used as input to the [BDT](#). They can be categorised into four groups: (i) positions and momenta of the lepton candidates, (ii) isolation variables, (iii) variables associated to the closest jet, and (iv) a quality variable that is specific to electron or muon candidate. The version of TOP LEPTONMVA used by this analysis is the same as [[21](#)], where a detailed description of all input features can be found.

Prompt leptons from $t\bar{t}W$, $t\bar{t}Z$, and tZq samples are used as signals in the **BDT** training while *nonprompt* leptons from $t\bar{t}$ samples are used as backgrounds. The trained **BDT** outputs a single score for each lepton candidate ranging from -1 to 1 with -1 (1) being the most background- (signal-) like. The tight working point with a threshold of ($>$) 0.9 is chosen as the selection criteria for both electron candidates and muon candidates, which corresponds to a signal(background) efficiency of 90%(1%). The strategy is to trade a small percentage ($<10\%$) of signal efficiency for several factors of background rejection.

13.1.2 Full Selection

In addition to the TOP LEPTONMVA requirement, a set of common selection criteria are applied to both electron and muon candidates. The minimum p_T requirement is 38 GeV, 20 GeV, and 20 GeV for the leading, sub-leading, and trailing lepton in p_T , respectively. This requirement is driven by the p_T thresholds of the **HLT** triggers to avoid inefficiency at turn-on. Electron and muon candidates are required to be in the pseudorapidity range $|\eta| < 2.4$, which corresponds to the acceptance of **CMS** tracker and muon system in 2016-2018. The transverse (longitudinal) impact parameters with respect to the **PV**, denoted as d_{xy} (d_z), is required to be in the range $|d_{xy}| < 0.05$ cm ($|d_z| < 0.05$ cm). The significance of the 3-dimensional impact parameter, denoted as SIP_3 , is defined as the 3-dimensional impact parameter divided by its uncertainty. It is required that $SIP_3 < 8$. The three cuts on impact parameters are added due to the difference in distributions of these parameters between *prompt* and *nonprompt* leptons. Also, they are part of the pre-selection requirement in the **BDT** training.

Furthermore, all lepton candidates are required to be isolated. This is achieved by first defining a cone with a distance parameter of ΔR around each lepton candidate, where $\Delta R = \sqrt{\Delta\eta^2 + \Delta\phi^2}$. Only particles within $\Delta R < R_{\max}$ can contribute to the isolation variable, where R_{\max} is referred to as the size of the cone. Secondly, a particle flow (**PF**) based isolation variable is defined as,

$$I_{\min}^{\text{rel}} = \frac{1}{p_T^\ell} \left\{ \sum_{\text{charged}} p_T + \max(0, \sum_{\text{neutral}} p_T - \rho \mathcal{A} \left[\frac{\Delta R}{0.3} \right]^2) \right\}, \quad (13.1)$$

where p_T^ℓ is the p_T of the lepton candidate, the first term inside the curly braces is the scalar sum of all charged particles associated with the **PV** while the second term evaluates the contribution from neutral particles. This is done by first scalar-summing over p_T of all neutral particles associated to the **PV**. A correction term, known as effective area correction [23], is then subtracted. This term is used to mitigate the impact of pile-up (**PU**) interactions. The size of the cone scales with p_T^ℓ as,

$$R_{\max} = \max(0.05, \min(0.2, \frac{10\text{GeV}}{p_T^\ell})). \quad (13.2)$$

This type of isolation variable is known as “mini” isolation, which maximises the signal efficiency at p_T^ℓ by reducing the cone size. It is required that lepton candidates to have $I_{\text{mini}}^{\text{rel}} < 0.12$.

For electron specifically, candidates are required to have a GSF track with one of less missing inner hit. Electron candidates reconstructed in the transition region between ECAL barrel and endcap (i.e. $1.44 < |\eta_{\text{SC}}| < 1.57$) are removed from consideration. For muon specifically, candidates are required to be **PF** muons and pass the medium working point discussed in [section 8.2](#).

Lepton candidates that pass all requirements stated above are referred to as “*tight*” leptons. Leptons selected with a separate set of criteria, known as “*loose*”, is used in estimating the *nonprompt* background, which is discussed in [chapter 15](#). Unless explicitly stated, all lepton objects presented in this search are *tight* leptons.

The energy of electron candidates are calibrated through. The moment scale of muon candidates are calibrated for muon candidates with $p_T < 200$ GeV. Scale factors are applied to *tight* leptons to correct for the differences in reconstruction, isolation, and identification between data and **MC**. These scale factors are obtained using dilepton events in Z resonance window.

13.2 Jet Selection

Jet candidates are reconstructed from **PF** candidates using the anti- k_T algorithm described in [section 13.2](#) with a cone size of 0.4. Charged hadrons that are not associated to the **PV** are removed. Jet candidates are required to have a minimum p_T of 30 GeV and in in the pseudorapidity range $|\eta| < 2.4$, where b-tagging are still effective. It is further required that all jet candidates to be isolated from *tight* leptons. A cone of the size 0.4 around each jet candidate is defined and candidates will be removed if any *tight* leptons are found within such a cone. This procedure is implemented to remove the overlap between leptons and jets.

The two primary sources of background are (i) detector noise, and (ii) jets from **PU** interactions. To suppress detector noise, a set of cut-based selections are applied to jet candidates. This algorithm utilizes information from **PF** candidates, including: (i) fraction of charged (neutral) hadrons energy, (ii) fraction of charged (neutral) EM energy, (iii) fraction of muon energy, and (iv) object multiplicity. The “tightLepVeto” working point is chosen to select jet candidates, which corresponds to 98-99% signal efficiency.

The second algorithm is designed to reject jet candidates that originate from **PU** interactions. This algorithm is based on a **BDT** that utilizes: (i) the trajectories of tracks associated to the jets, (ii) the topology of the jet shape, and (iii) object multiplicity. The *loose* working point is chosen to select jet candidates with $p_T < 50$ GeV, which corresponds to 99% signal efficiency. Applying this algorithm to jet candidates with $p_T > 50$ GeV is both ineffective and unnecessary as **PU** jets mostly reside in low p_T spectrum. The overall effect of this algorithm on this analysis is small as **PU** jets constitute only a small fraction of all jet candidates in the phase space of this analysis.

As discussed in [section 8.4](#), the energy scale for all jet candidates (data and [MC](#)) are calibrated. One extra correction is applied to simulated jets to recreate the jet energy resolution as measured in data.

13.3 Identification of b jets

The DEEPJET algorithm [\[24\]](#) is used to identify jets that originate from b quark. The core strategy of this algorithm is to minimise information. This is achieved by removing entirely the selection of jet constituents, which limits the number of jet constituents considered. Additionally, an effort is made to use as many low-level features as possible, which further further deepens the feature space. Approximately a total of 650 input features are used, which can be categorized into four groups: (i) global variables, (ii) charged PF candidate features, (iii) neutral PF candidate features, (iv) and secondary vertex ([SV](#)) features associated with the jet. When compared to the existing DEEPCSV algorithm [\[25\]](#), DEEPJET algorithm delivers up to 20% improvement in signal efficiency while maintaining the same background efficiency.

The DEEPJET algorithm outputs a score ranging between 0 and 1, with 0 (1) being the most background- (signal-) like. The medium working point is chosen to tag b jet candidate, which corresponds to 70%-80% signal efficiency. The shape of the DEEPJET output distribution is corrected for the differences between data and [MC](#) in signal and background efficiencies. The per-event correction weight ω is defined as,

$$\omega = \prod_i^{N_{\text{jets}}} \text{SF}(p_{T_i}, \eta_i, F_i, D_i), \quad (13.3)$$

where SF is the ratio of efficiency in data to efficiency to [MC](#) parameterized as a function of p_T , η , (MC truth) flavor F, as well as DEEPJET output D of each jet in the event. ω is applied to all [MC](#) events, and additional scale factors are also applied to remove the normlization effect of ω . These scale factors are measured using [MC](#) in $e\mu\ell$ channel described in [chapter 14](#). The effect of these scale factors are shown in [Figure 13.1](#).

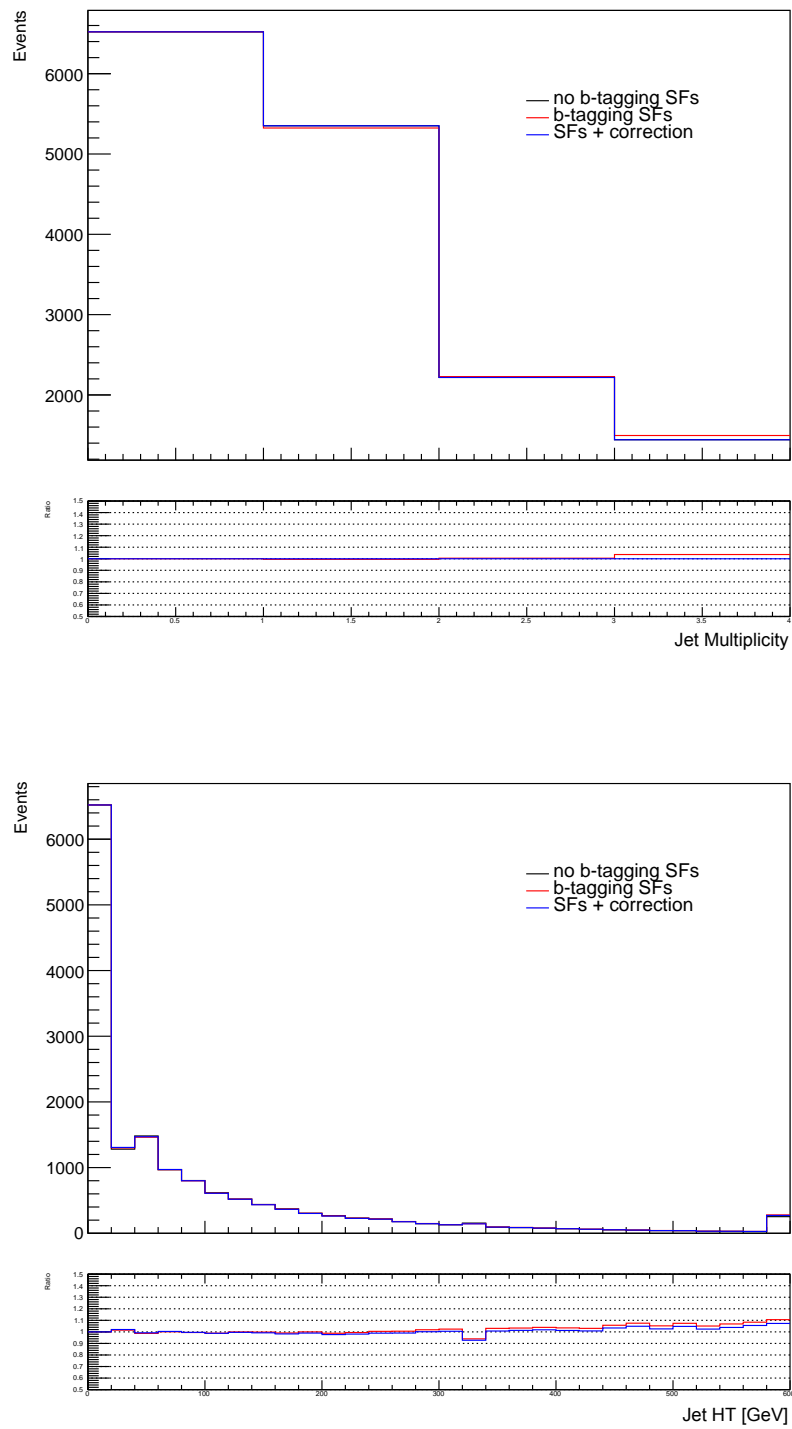


Figure 13.1: Simulated events in $e\mu\ell$ channel: jet multiplicity (top) and H_T (bottom).

CHAPTER 14

Event Selection

Events are required to contain exactly three *tight* leptons described in [section 13.1](#). Furthermore, events are selected with [HLT](#) triggers discussed in [section 12.3](#). Events with different lepton flavor composites are further categorized into three exclusive channels: eee , $\mu\mu\mu$, $e\mu\ell$. In all three channels, the sum of the electric charges of the selected leptons are required to be 1 or -1. The leading leptons in all selected events are required to be matched with trigger objects within $\Delta R < 0.2$. Within each channel, different regions are defined to further understand signal and background.

$e\mu\ell$ is the channel where close to 100% of the simulated signal events reside. This channel is divided into signal-enriched [SRs](#) and signal-depleted [VRs](#), which are discussed in [section 14.1](#) and [section 14.2](#), respectively. Due to the lack of different flavors, the eee and $\mu\mu\mu$ channels are signal-depleted by definition. Therefore, events found in these two channels are only used to study background processes, which are discussed in [section 14.2](#). The kinematic reconstruction of heavy particles, such as the top quark, is described in [section 14.3](#).

14.1 Signal Region

The core feature of the signal events is the presence of the “LFV $e\mu$ ” pair, which consists of a pair of opposite-sign and different-flavor ([OSDF](#)) leptons. It is guaranteed that there is at least one [OSDF](#) pair in all events residing in $e\mu\ell$ channel due to the requirement on electric charges. The [OSDF](#) pair is immediately labelled as the LFV $e\mu$ pair if it is only possible to reconstruct one [OSDF](#) pair. In events where a pair of same-sign and same-flavor ([SSSF](#)) leptons are present, a kinematic reconstruction is used to determine which one of two leptons form the LFV $e\mu$ pair with the third lepton, which is detailed in [section 14.3](#). Leptons that form the LFV $e\mu$ pair are referred to as the LFV electron or muon as it is assumed that they originate from the [CLFV](#) interaction. Based on the event topology of the signal process, further selection criteria are applied to define the [SR](#). These selection criteria help achieve an optimal signal to background ratio by removing majority of the background events present in $e\mu\ell$ channel.

At tree-level, signal events are expected to contain one or two jets, which motivates a requirement of at least one jet in [SR](#). Furthermore, it is required that there is no more than one b-tagged

jet to suppress the contribution from $t\bar{t}$ events. Another prominent background is Drell-Yan production that features an $OSSF$ pair. To suppress Drell-Yan processes in SR , events that contain an $OSSF$ lepton pair with an invariant mass between 50 GeV and 106 GeV are removed. The lower bound of this veto is lower than the typical value (e.g. 75 GeV) because the mass range between 50 GeV and 75 GeV has very few signal events and is dominated by *nonprompt* background from photon conversion. Additionally, a modest threshold of 20 GeV is applied to MET due to the presence of neutrinos in the signal events.

Distributions of the LFV $e\mu$ mass and the Z boson mass are shown in Figure 14.1. All backgrounds in Figure 14.1 are estimated using MC simulation even though strategy is to use a data-driven method to estimate the *nonprompt* background. This serves as a preliminary check to understand the components of different backgrounds in SR . Distributions of more variables in SR are included in Appendix B.

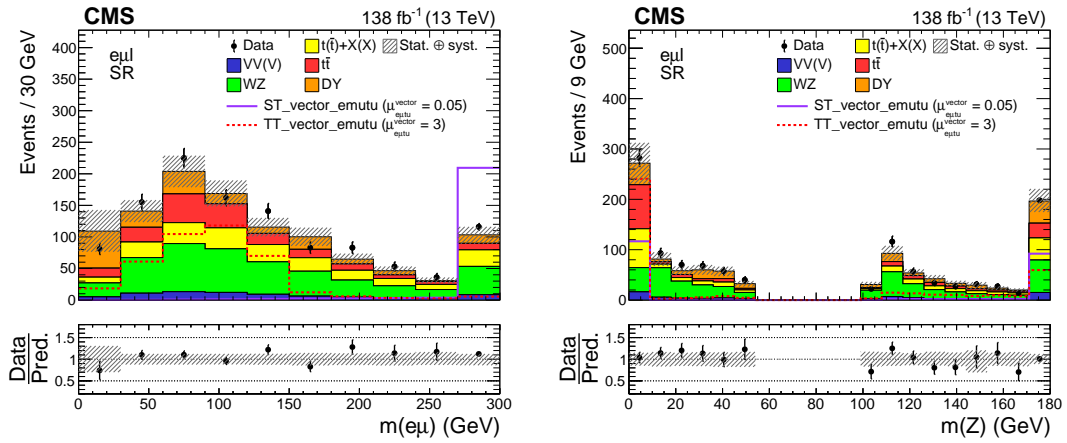


Figure 14.1: Distributions of the LFV $e\mu$ mass (left) and the Z boson mass (right) in SR . The data are shown as filled points and the SM background predictions as histograms. The $VV(V)$ background includes ZZ and triboson production, while the $t\bar{t} + X(X)$ component includes $t\bar{t}W$, $t\bar{t}Z$, $t\bar{t}H$, tZq , and smaller backgrounds containing one or two top quarks plus a boson or quark. All backgrounds are estimated using MC simulation. The hatched bands indicate statistical and systematic uncertainties for the SM background predictions. The normalisation of the signal processes is chosen arbitrarily for improved visualisation. The last bin of both histograms includes the overflow events.

Using the LFV $e\mu$ mass, the SR is further divided into two subsets to create top production and decay enriched regions:

- $SR1$, $m_{e\mu} < 150$ GeV: top decay enriched.
- $SR2$, $m_{e\mu} > 150$ GeV: top production enriched.

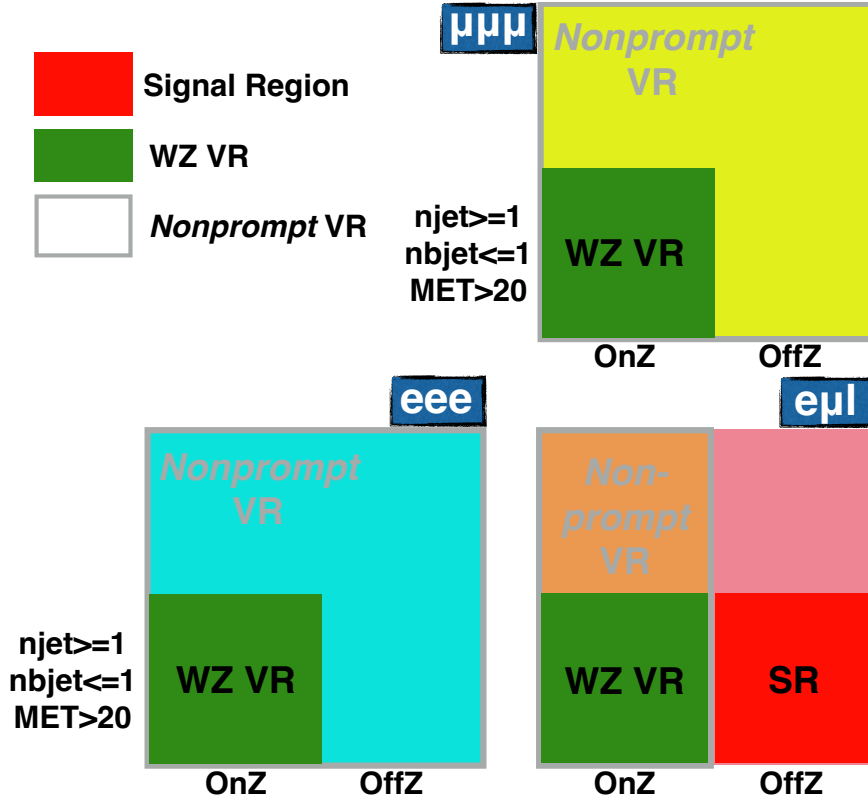


Figure 14.2: Illustration of selection criteria used to define different regions. “OnZ” means the presence of at least one $OSSF$ pair with an invariant mass between 50 GeV and 106 GeV. Events are labelled as “OffZ” when they fail “OnZ” criteria.

14.2 Validation Region

There are two types of signal-depleted **VR** defined across three channels: *nonprompt VR* and *WZ VR*. The purpose of these two types of **VR** is only limited to the validation of the background modelling as neither of them enter the final fit. It is expected that the *nonprompt VR* has a significant fraction of *nonprompt* background while *WZ* production is responsible for most of the backgrounds in the *WZ VR*. Distributions of leading lepton p_T and leading lepton η in *WZ* control region can be found in Figure 14.3. The *nonprompt VR*s are further discussed in chapter 15.

Selection criteria used to define different regions are illustrated in Figure 14.2 and is summarized in Table 14.1.

14.3 Kinematic Reconstruction

As mentioned, the LFV $e\mu$ pair is assumed to be the product of the $CLFV$ interaction, while the third lepton, referred to as the standalone lepton, is assumed to originate from the leptonically

Channel	Region	OnZ	OffZ	MET > 20 GeV	njet ≥ 1	nbjet ≤ 1
eee	VR	-	-	-	-	-
	WZ VR	✓	-	✓	✓	✓
eμℓ	SR	-	✓	✓	✓	✓
	Nonprompt VR	✓	-	-	-	-
	WZ VR	✓	-	✓	✓	✓
μμμ	Nonprompt VR	-	-	-	-	-
	WZ VR	-	✓	✓	✓	✓

Table 14.1: Summary of the selection criteria used to define different event regions. “OnZ” means the presence of at least one **OSSF** pair with an invariant mass between 50 GeV and 106 GeV. Events are labelled as “OffZ” when they fail “OnZ” criteria.

decaying top quark. To distinguish this top quark ($t \rightarrow \ell \nu b$) with the top quark that decays via the **CLFV** interaction ($t \rightarrow e \mu q$), the former is referred to as the **SM** top quark while the latter is referred to as the LFV top quark.

Jet with the highest b-tagging score, regardless of whether or not it crosses the medium working point threshold, is assumed to originate from bottom quark decay. Therefore, it is combined with **MET** and to build the **SM** top quark. The x and y component of **MET** are taken as measurements of neutrino p_x and p_y . The z component of neutrino momentum is calculated by imposing the constraint that the invariant mass of the combined object (standalone lepton + neutrino) must be equal to W boson mass. If there is no real solution, the real part of the complex solution is taken. If there is more than one real solution, the solution that is the closest to the p_z of the standalone lepton is taken. In events where there is more than one candidate of standalone lepton (i.e. the presence of the **SSSF** pair), the lepton that gives a top mass that is the closest to the **SM** top quark mass ($m_t = 172.5$ GeV) is taken as the standalone lepton.

Once the standalone lepton has been determined, the remaining two leptons are labelled as the LFV eμ pair and is combined with each selected jet to reconstruct the LFV top quark candidates. Jet with the highest b-tagging score is excluded from this reconstruction since it is assumed to be from the decay of the **SM** top quark. Out of all the LFV top quark candidates, the candidate that gives a top mass that is the closest to the **SM** top quark mass is taken. The LFV top quark mass is set to 0 in events where there are less than two jets.

Z boson candidate is reconstructed using the **OSSF** pair, which is not guaranteed to be present in the eμℓ channel. The Z boson mass (m_Z) is set to 0 in events where the **OSSF** is absent. Z boson candidate is the only heavy particle reconstructed in the eee and μμμ channels. Since

there are always two ways to form the **OSSF** pair, the **OSSF** pair with an invariant mass that is closer to the Z boson mass ($m_Z = 91.2 \text{ GeV}$) is taken.

Jets with high b-tagging scores are combined with leptons to form so called “ $m_{b\ell}$ ” systems. The first $m_{b\ell}$ system takes the jet with the highest b-tagging score and combine it with each *tight* lepton in events. Out of the three $m_{b\ell}$ system candidates, the one with the lowest $m_{b\ell}$ is taken, the two constitutes are excluded from the consideration of the second $m_{b\ell}$ system. If additional jets exist, the second $m_{b\ell}$ system takes jet with the highest b-tagging score and combine it with two of the remaining leptons separately. Out of the two candidates, the one with lowest $m_{b\ell}$ is taken. $m_{b\ell}$ is set to 0 if no additional jet exists after the formation of the first $m_{b\ell}$ system.

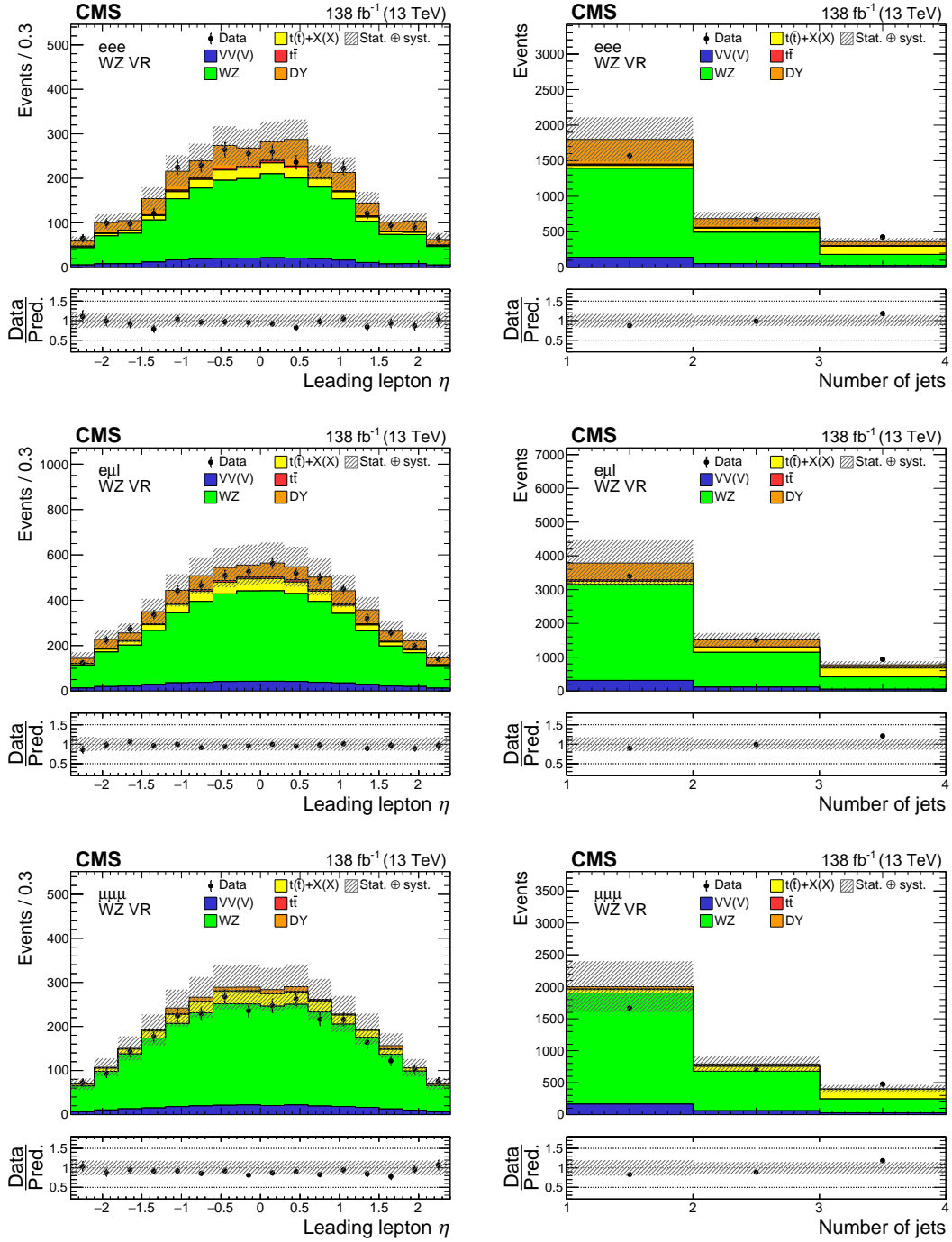


Figure 14.3: Distributions of the leading lepton η (left column) and the jet multiplicity (right column) in the WZ VRs. Events in the eee, $e\mu\ell$, and $\mu\mu\mu$ WZ VRs are shown in the upper, middle, and lower row, respectively. The data are shown as filled points and the background predictions as histograms. All backgrounds are estimated with MC simulation. The hatched bands indicate statistical and systematic uncertainties for the background predictions. The last bin of the right column histograms includes the overflow events.

CHAPTER 15

Nonprompt Background Estimation

In this analysis, the term *prompt* leptons refers to leptons that originate from the CLFV vertex, the Drell-Yan process, or an electroweak boson decay, including leptons from τ decays if the τ lepton originates from the latter two processes. *Nonprompt* leptons refer to leptons that originate from hadron decays and photon conversions, as well as particles misidentified as leptons. *Nonprompt* leptons are suppressed through isolation requirements and a multivariate analysis (MVA)-based identification specifically trained to reject them.

Nonprompt backgrounds are defined to be backgrounds with at least one *nonprompt* lepton passing the *tight* selection criteria, in this case generally dominated by Drell-Yan and $t\bar{t}$ production. An accurate estimation of *nonprompt* backgrounds is difficult to achieve through MC modelling. Therefore, a data-driven technique called the “*matrix method*” [26] is used to estimate the *nonprompt* backgrounds.

A brief description of the *matrix method* in its simplest form is given in section 15.1 followed by its generalization and implementation in section 15.2. This method is validated using three VRs and is described in section 14.2. Lastly, the *nonprompt* estimation in the SR is presented in section 15.4.

15.1 The Matrix Method

The *matrix method* is a data driven technique used to estimate the fraction of *nonprompt* leptons that pass a given lepton selection, referred to as “*tight*”. The *tight* selection usually incorporates tight lepton identification and isolation requirements and corresponds to the full lepton selection used in an analysis. The *loose* selection is obtained by loosening the *tight* selection. The *loose* selection is used as a baseline such that any *loose* leptons fall into one of the two exclusive categories: *tight* or *not tight*. The *matrix method* deals with *prompt* and *nonprompt* leptons separately. As a result, *prompt* and *nonprompt* efficiencies are introduced, as is illustrated in Figure 15.1.

In a simplified scenario with only one lepton in the final state, the *prompt* efficiency r measures the probability of *prompt* leptons pass *tight* selection. It is treated as an observable that can be obtained through measurement,

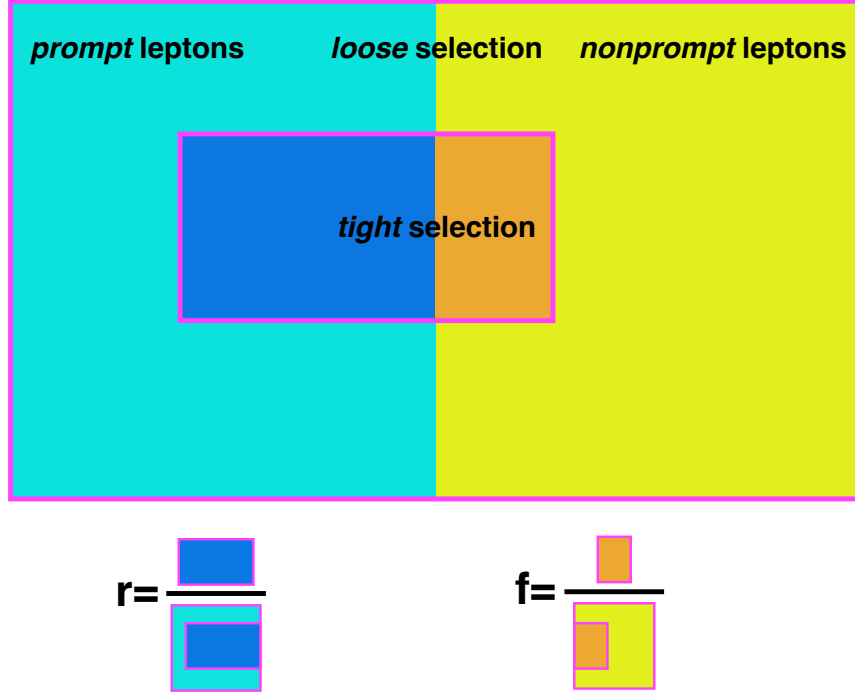


Figure 15.1: Illustration of the *prompt* efficiency r and the *nonprompt* efficiency f .

$$r = \frac{n_P^T}{n_P^T + n_P^{\bar{T}}}, \quad (15.1)$$

in which $n_P^T/n_P^{\bar{T}}$ denotes the number of events with a *prompt* lepton that is *tight/not tight*.

Similarly, *nonprompt* efficiency f can be expressed as,

$$f = \frac{n_N^T}{n_N^T + n_N^{\bar{T}}}, \quad (15.2)$$

in which $n_N^T/n_N^{\bar{T}}$ denotes the number of events with a *nonprompt* lepton that is *tight/not tight*.

The measurement of r/f is often performed in dedicated control regions, where high purity of *prompt/nonprompt* leptons is expected. These regions are referred to as the **MR**. It is assumed that r/f is a universal property of *prompt/nonprompt* leptons that is independent of physics processes. Therefore, r/f extracted from **MR** can be used to estimate the contamination of *nonprompt* leptons in a different region (e.g. **SR**) even though these two regions are orthogonal to each other.

In this simplified scenario, the total number of events in the region of interest (e.g. **SR/VR**) with a *tight/not tight* lepton can be expressed in a system of equations,

$$\begin{aligned} N^T &= N_P^T + N_N^T \\ N^{\bar{T}} &= N_P^{\bar{T}} + N_N^{\bar{T}}, \end{aligned} \quad (15.3)$$

in which capital letter “ N ” is used to indicate that these numbers are referring to events in a region that is different from MR. $N_P^{\bar{T}}/N_N^{\bar{T}}$ can be expressed in terms of r/f and N_P^T/N_N^T according to Equation 15.1/15.2 and the assumption that r/f remains the same across different regions,

$$\begin{aligned} N^T &= r \frac{N_P^T}{r} + f \frac{N_N^T}{f} \\ N^{\bar{T}} &= (1-r) \frac{N_P^T}{r} + (1-f) \frac{N_N^T}{f}. \end{aligned} \quad (15.4)$$

Equation 15.4 can also be expressed in the form of matrix,

$$\begin{pmatrix} N^T \\ N^{\bar{T}} \end{pmatrix} = \begin{pmatrix} r & f \\ 1-r & 1-f \end{pmatrix} \begin{pmatrix} N_P^T/r \\ N_N^T/f \end{pmatrix}. \quad (15.5)$$

Regions that correspond to the two numbers that appear in the righthand side vector of Equation 15.5 are referred to as the “application regions (ARs)”, which can be constructed using experimental data. The estimation of *nonprompt* background, denoted by N_N^T , can be obtained by a simple matrix inversion.

15.2 Generalization and Implementation of the Matrix Method

The description in previous section deals with a scenario where only one lepton is studied. This analysis uses a generalized version of the *matrix method*, where all three *tight* leptons are considered to be possibly *nonprompt*. Equation 15.5 is generalized as,

$$\begin{pmatrix} N^{TTTT} \\ N^{TTT\bar{t}} \\ N^{TT\bar{t}t} \\ N^{T\bar{t}t\bar{t}} \\ N^{\bar{t}t\bar{t}t} \\ N^{\bar{t}t\bar{t}\bar{t}} \\ N^{\bar{t}t\bar{t}t} \\ N^{\bar{t}t\bar{t}\bar{t}} \end{pmatrix} = \begin{pmatrix} r_1 r_2 r_3 & r_1 r_2 \bar{f}_3 & r_1 \bar{f}_2 r_3 & r_1 \bar{f}_2 \bar{f}_3 & \bar{f}_1 r_2 r_3 & \bar{f}_1 r_2 \bar{f}_3 & \bar{f}_1 \bar{f}_2 r_3 & \bar{f}_1 \bar{f}_2 \bar{f}_3 \\ r_1 r_2 (1-r_3) & r_1 r_2 (1-\bar{f}_3) & r_1 \bar{f}_2 (1-r_3) & r_1 \bar{f}_2 (1-\bar{f}_3) & \bar{f}_1 r_2 (1-r_3) & \bar{f}_1 r_2 (1-\bar{f}_3) & \bar{f}_1 \bar{f}_2 (1-r_3) & \bar{f}_1 \bar{f}_2 (1-\bar{f}_3) \\ r_1 (1-r_2) r_3 & r_1 (1-r_2) \bar{f}_3 & r_1 (1-\bar{f}_2) r_3 & r_1 (1-\bar{f}_2) \bar{f}_3 & \bar{f}_1 (1-r_2) r_3 & \bar{f}_1 (1-r_2) \bar{f}_3 & \bar{f}_1 (1-\bar{f}_2) r_3 & \bar{f}_1 (1-\bar{f}_2) \bar{f}_3 \\ r_1 (1-r_2)(1-r_3) & r_1 (1-r_2)(1-\bar{f}_3) & r_1 (1-\bar{f}_2)(1-r_3) & r_1 (1-\bar{f}_2)(1-\bar{f}_3) & \bar{f}_1 (1-r_2)(1-r_3) & \bar{f}_1 (1-r_2)(1-\bar{f}_3) & \bar{f}_1 (1-\bar{f}_2)(1-r_3) & \bar{f}_1 (1-\bar{f}_2)(1-\bar{f}_3) \\ (1-r_1) r_2 r_3 & (1-r_1) r_2 \bar{f}_3 & (1-r_1) \bar{f}_2 r_3 & (1-r_1) \bar{f}_2 \bar{f}_3 & (1-\bar{f}_1) r_2 r_3 & (1-\bar{f}_1) r_2 \bar{f}_3 & (1-\bar{f}_1) \bar{f}_2 r_3 & (1-\bar{f}_1) \bar{f}_2 \bar{f}_3 \\ (1-r_1) r_2 (1-r_3) & (1-r_1) r_2 (1-\bar{f}_3) & (1-r_1) \bar{f}_2 (1-r_3) & (1-r_1) \bar{f}_2 (1-\bar{f}_3) & (1-\bar{f}_1) r_2 (1-r_3) & (1-\bar{f}_1) r_2 (1-\bar{f}_3) & (1-\bar{f}_1) \bar{f}_2 (1-r_3) & (1-\bar{f}_1) \bar{f}_2 (1-\bar{f}_3) \\ (1-r_1)(1-r_2) r_3 & (1-r_1)(1-r_2) \bar{f}_3 & (1-r_1)(1-\bar{f}_2) r_3 & (1-r_1)(1-\bar{f}_2) \bar{f}_3 & (1-\bar{f}_1)(1-r_2) r_3 & (1-\bar{f}_1)(1-r_2) \bar{f}_3 & (1-\bar{f}_1)(1-\bar{f}_2) r_3 & (1-\bar{f}_1)(1-\bar{f}_2) \bar{f}_3 \\ (1-r_1)(1-r_2)(1-r_3) & (1-r_1)(1-r_2)(1-\bar{f}_3) & (1-r_1)(1-\bar{f}_2)(1-r_3) & (1-r_1)(1-\bar{f}_2)(1-\bar{f}_3) & (1-\bar{f}_1)(1-r_2)(1-r_3) & (1-\bar{f}_1)(1-r_2)(1-\bar{f}_3) & (1-\bar{f}_1)(1-\bar{f}_2)(1-r_3) & (1-\bar{f}_1)(1-\bar{f}_2)(1-\bar{f}_3) \end{pmatrix} \begin{pmatrix} N_{PPP}^{TTTT}/r_1 r_2 r_3 \\ N_{PPN}^{TTTT}/r_1 r_2 \bar{f}_3 \\ N_{PNP}^{TTTT}/r_1 \bar{f}_2 r_3 \\ N_{PNN}^{TTTT}/r_1 \bar{f}_2 \bar{f}_3 \\ N_{NPP}^{TTTT}/\bar{f}_1 r_2 r_3 \\ N_{NPN}^{TTTT}/\bar{f}_1 r_2 \bar{f}_3 \\ N_{NNP}^{TTTT}/\bar{f}_1 \bar{f}_2 r_3 \\ N_{NNN}^{TTTT}/\bar{f}_1 \bar{f}_2 \bar{f}_3 \end{pmatrix}. \quad (15.6)$$

All but the first number that appear in the righthand side vector correspond to events with at least one *nonprompt* lepton that pass *tight* selection criteria. Therefore, the overall *nonprompt* background is expressed as,

$$N_{Nonprompt}^{TTTT} = N_{PPN}^{TTTT} + N_{PNP}^{TTTT} + N_{PNN}^{TTTT} + N_{NPP}^{TTTT} + N_{NPN}^{TTTT} + N_{NNP}^{TTTT} + N_{NNN}^{TTTT}, \quad (15.7)$$

which can be obtained by first constructing 8 **ARs** to form the lefthand side vector. Secondly, the 8×8 matrix is constructed and inverted. Lastly, the righthand side vector can be obtained by multiplying the lefthand side vector to the inverted matrix.

Only two **PDs** “SingleElectron” and “SingleMuon” are used in the construction of **MR** in 2016 and 2017 while “SingleElectron” is replaced with “EGamma” in 2018. In addition to **PDs**, the measurements of r/f also utilize the $t\bar{t}$ sample and all **MC** samples listed under the “prompt background” category in Table 12.3. Depending on the flavor of the leading lepton in **MC**, events are selected with either a single-electron or a single-muon trigger, which is summarized in Table 15.1. Data events are selected with the same **HLT** triggers as well but events in “SingleMuon” (“SingleElectron” or “EGamma”) **PD** are only accepted if the leading lepton is a muon (electron).

Channel	Path	Dataset	2016	2017	2018
Electron	HLT_Ele27_WPTight_Gsf	Data & MC	✓	-	-
	HLT_Ele35_WPTight_Gsf	Data & MC	-	✓	✓
Muon	HLT_IsoMu27	Data & MC	✓	✓	✓

Table 15.1: Summary of the **HLT** triggers used in the measurement of r and f .

Both r and f are parameterized in bins of lepton p_T , $|\eta|$, and jet multiplicity. The bin range is optimized to retain sufficient statistics for each bin:

- Electron p_T bin range: $\{20.0, 24.6, 28.8, 33.0, 37.2, 41.4, 46.1, 52.1, 59.3, 68.3, 82.7, 110.6\}$ GeV,
- Muon p_T bin range: $\{20.0, 23.8, 27.7, 31.3, 35.0, 38.9, 42.8, 45.6, 50.7, 59.5, 72.9, 94.3\}$ GeV,
- $|\eta|$ bin range: $\{0, 0.8, 1.6, 2.4\}$,
- Jet multiplicity: $\{0 \text{ jet}, \geq 1 \text{ jet}\}$.

The jet multiplicity bin is a proxy for variation of the composition of physics processes. In addition to requiring at least one jet, the **MR** corresponds to the second jet multiplicity bin requires no more than one b-tagged jet as this is also required in the **SR**.

The *nonprompt* efficiency is measured in same-sign dilepton regions, in which the leading lepton in p_T , used as a *tag*, is required to be matched with trigger objects within $\Delta R < 0.2$. The sub-leading lepton is required to pass the *loose* selection and is taken as the *probe*. Events that have two same-sign electrons with an invariant mass between 76 GeV and 106 GeV are removed from **MR** to suppress the backgrounds that originate from charge misidentification. No such requirement has been introduced to the muon **MR** due to its negligible rate of charge misidentification.

The contribution from *prompt* backgrounds, estimated from **MC** simulation, are subtracted from data. A representative composition of backgrounds in **MR** is shown in Figure 15.2.

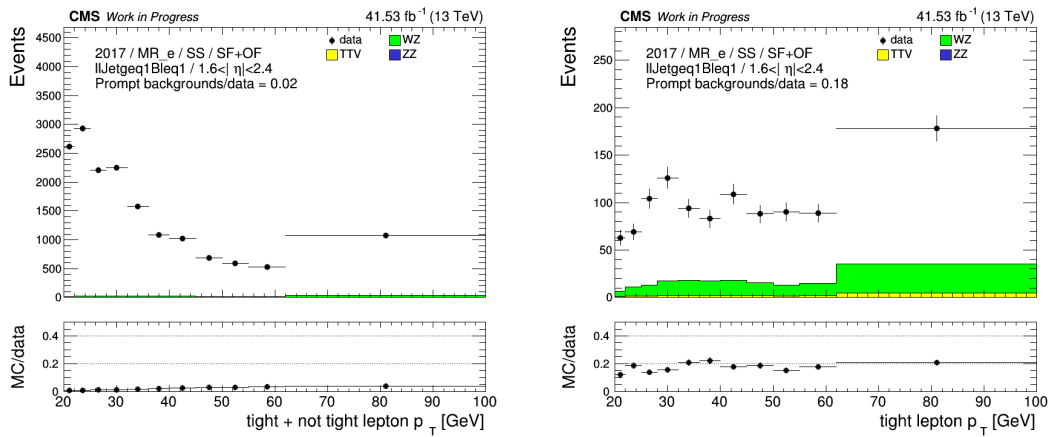


Figure 15.2: Distribution of lepton p_T in a representative electron *nonprompt* efficiency **MR**. In this particular example, both ee and μe flavor composites are considered. At least one jet and at most one b-tagged jet are required (the second jet multiplicity bin). *Probe* electron is required to have $1.6 < |\eta| < 2.4$ (the third η bin). Contamination from *prompt* backgrounds are estimated with **MC** simulation, and are shown as histograms. The data are shown as filled points. From left to right: *loose* (i.e. *tight + not tight*) electron p_T , *tight* electron p_T .

The fake efficiency f is calculated as:

$$f = \frac{n_{data}^{tag+tight} - n_{MC(prompt)}^{tag+tight}}{n_{data}^{tag+loose} - n_{MC(prompt)}^{tag+loose}}, \quad (15.8)$$

where the numerator is selected with one *tag* and one *tight* lepton while the denominator is selected with one *tag* and one *loose* lepton. The selection criteria for *tag*, *loose*, and *tight* lepton is summarised in Table 15.3.

Lepton	Selection	<i>loose</i>	<i>tag</i>	<i>tight</i> ⁱⁱ
Electron	p_T	> 20 GeV	> 38 GeV ⁱ	> 20 GeV
	I_{mini}^{rel}	<0.4	<0.1	<0.12
	TOP LEPTONMVA	>-0.9	>0.95	>0.9
	Match with trigger objects	-	✓	-
Muon	p_T	> 20 GeV	> 30 GeV	> 20 GeV
	I_{mini}^{rel}	<0.4	<0.1	<0.12
	Cut-based ID	-	Medium WP	Medium WP
	TOP LEPTONMVA	>0.5	>0.9	>0.9
	Match with trigger objects	-	✓	-

Table 15.2: Summary of the lepton selections needed for the measurement of r and f . Please note: (i) the minimum p_T cut for *tag* electron in 2016 dataset is reduced to 30 GeV to adjust for the trigger threshold, and (ii) the *tight* selection here is the same as the *tight* lepton selection described in section 13.1.

The measured *nonprompt* efficiency f exhibits a dependency on flavor composition, as is shown in Figure 15.3. This dependency is treated as a source of the systematic uncertainties of the *nonprompt* estimation and is further discussed in section 17.2.

The *prompt* efficiency r is measured in simulated $t\bar{t}$ events in opposite-sign dilepton regions. The same lepton selection listed in Table 15.2 is used to perform the *Tag-and-Probe*. The leading lepton in p_T is used as a *tag* while the oppositely charged sub-leading lepton is taken as a *probe*. The variation of r between different flavor composition is negligible, as is shown in Figure 15.4. Therefore, only $e\mu$ events are used to measure *prompt* efficiency in order to minimise the contamination of *nonprompt* leptons.

The selection criteria for various MRs is summarised in Table 15.3.

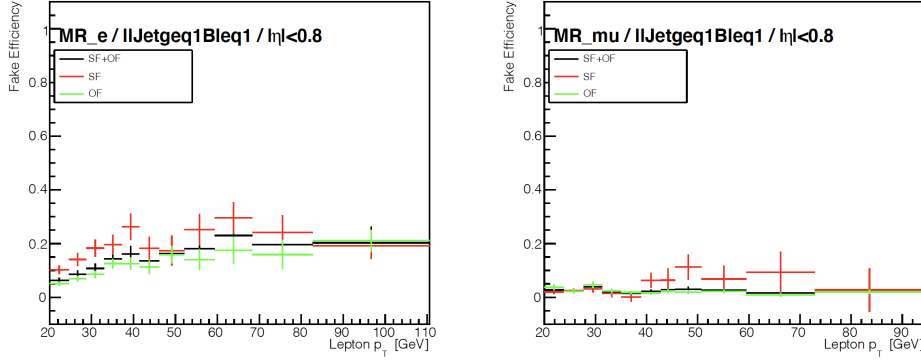


Figure 15.3: Representative *nonprompt* electron efficiency measured in data events. From left to right: electron f , muon f . Events with a same-flavor lepton pair are shown in red points while events selected with a different-flavor lepton pair are shown in green points. Events with a same-flavor or different-flavor lepton pair is shown in black points. These plots correspond to the first $|\eta|$ bin ($|\eta| < 0.8$) and the second jet multiplicity bin. Events selected Error bars displayed in these plots include statistical uncertainty only.

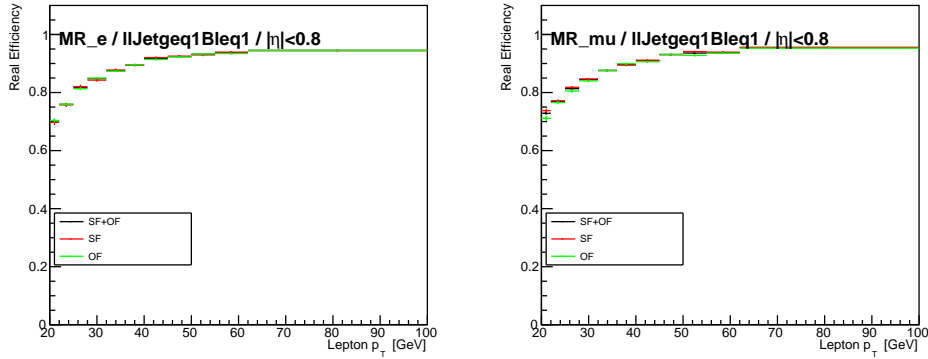


Figure 15.4: Representative *prompt* efficiency measured in simulated $t\bar{t}$ events. From left to right: electron r , muon r . Events with a same-flavor lepton pair are shown in red points while events selected with a different-flavor lepton pair are shown in green points. Events with a same-flavor or different-flavor lepton pair is shown in black points. These plots correspond to the first $|\eta|$ bin ($|\eta| < 0.8$) and the second jet multiplicity bin. Error bars displayed in these plots include statistical uncertainty only.

15.3 Validation of the Matrix Method

The performance of the *matrix method* is validated using three regions that are tangential to the **SR**, referred to as **VRs**. In these **VRs**, *prompt* backgrounds are estimated using **MC** simulation while *nonprompt* background is estimated with the *matrix method*. A summary of the selections

Observable	jet bin	# of selected leptons	lepton flavor composite	$ \sum_i C_i $	OffZ	njet	nbjet
f	0 jet	2	any	2	same-sign ee	= 0	= 0
	1 or more jet	2	any	2	same-sign ee	≥ 1	≤ 1
r	0 jet	2	e μ only	0	-	= 0	= 0
	1 or more jet	2	e μ only	0	-	≥ 1	≤ 1

Table 15.3: Summary of the cuts applied to the r/f measurement region.

applied to these VRs is given in [chapter 14](#).

Distribution of the leading lepton η and jet multiplicity are shown in [Figure 15.5](#). Good agreement between data and background estimate has been observed in all three [VRs](#).

15.4 Nonprompt Estimate in SR

The *matrix method* is used to estimate *nonprompt* background in the [SR](#). Distributions of the LFV e μ mass and the Z boson mass are shown in [Figure 15.6](#). When compared to background estimate from pure [MC](#) simulation ([Figure 14.1](#)), the updated background template is smoother with lower statistical uncertainties.

The number of expected events from various kinds of backgrounds are shown in [Table 15.4](#).

Table 15.4: Expected background contributions and the number of events observed in data collected during 2016–2018. The statistical and systematic uncertainties are added in quadrature. The category “other backgrounds” includes smaller background contributions containing one or two top quarks plus a boson or quark. The **CLFV** signal, generated with $C_{\text{e}\mu\text{t}\text{u}}^{\text{vector}}/\Lambda^2 = 1\text{TeV}^{-2}$, is also listed for reference. The signal yields include contributions from both top production and decay modes.

Process	$m(\text{e}\mu) < 150 \text{ GeV}$	$m(\text{e}\mu) > 150 \text{ GeV}$
Nonprompt	351 ± 92	146 ± 38
WZ	275 ± 64	145 ± 35
ZZ	33.2 ± 6.5	13.1 ± 2.6
VVV	17.0 ± 8.5	12.0 ± 6.0
$t\bar{t}W$	47.6 ± 10.0	40.0 ± 9.1
$t\bar{t}Z$	39.1 ± 7.9	25.8 ± 5.4
$t\bar{t}H$	28.2 ± 4.5	10.0 ± 1.6
tZq	5.5 ± 1.1	2.5 ± 0.5
Other	7.3 ± 3.7	4.5 ± 2.3
Total expected	805 ± 123	398 ± 57
Data	783	378
CLFV	207 ± 15	4440 ± 215

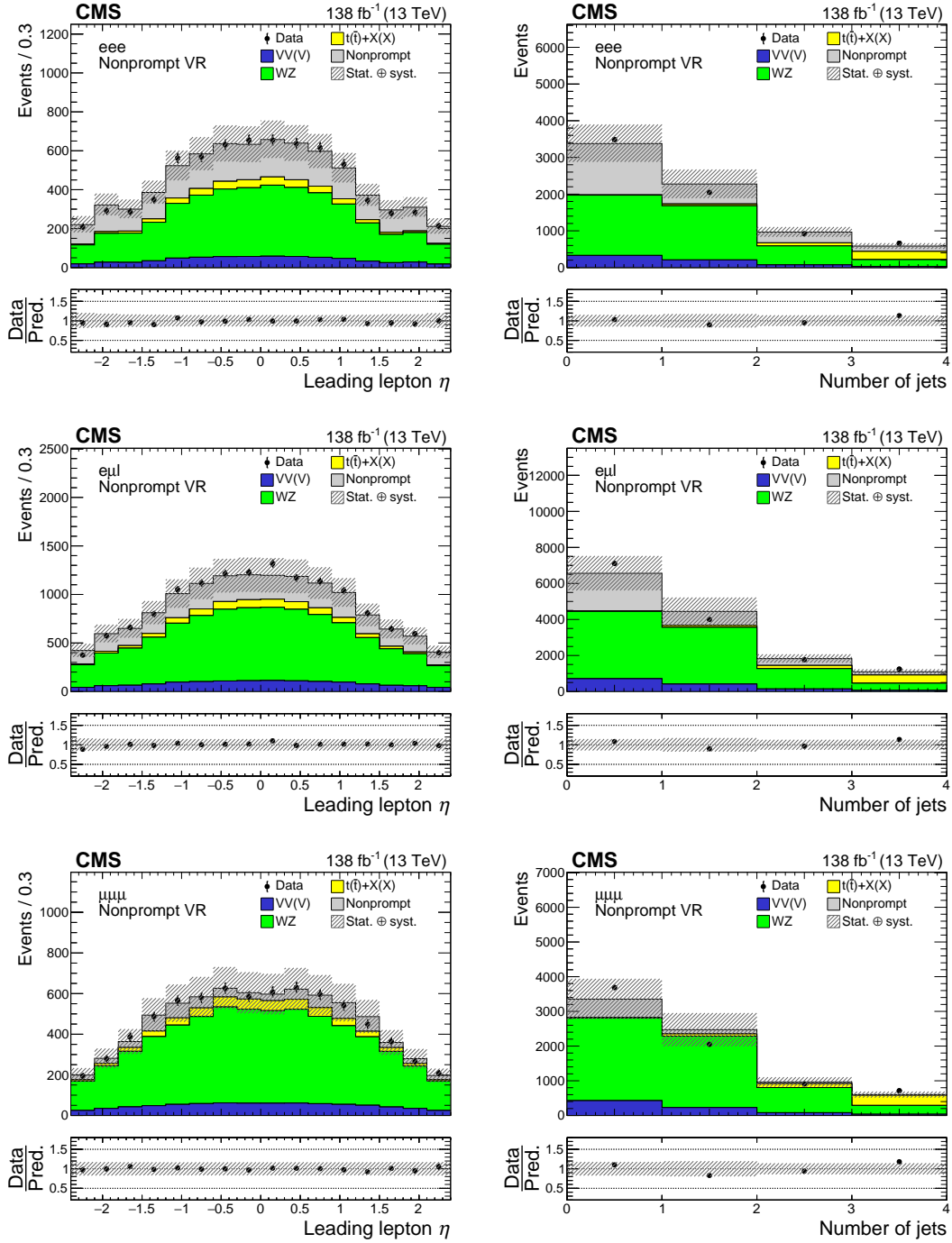


Figure 15.5: Distributions of the leading lepton η (left column) and the jet multiplicity (right column) in the *nonprompt VRs*. Events in the eee , $e\mu\ell$, and $\mu\mu\mu$ *nonprompt VRs* are shown in the upper, middle, and lower row, respectively. The data are shown as filled points and the **SM** background predictions as histograms. The *nonprompt* background is estimated using control samples in data, while other backgrounds are estimated using **MC** simulation. The hatched bands indicate statistical and systematic uncertainties for the **SM** background predictions. The last bin of the right histogram includes the overflow events.

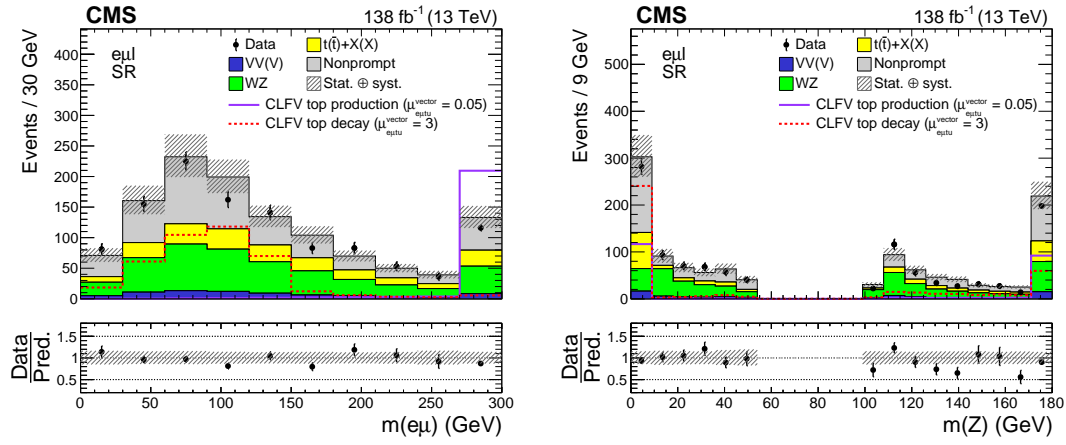


Figure 15.6: Distributions of the LFV $e\mu$ mass (left) and the Z boson mass (right) in **SR**. The data are shown as filled points and the **SM** background predictions as histograms. The *nonprompt* background is estimated using control samples in data, while other backgrounds are estimated using **MC** simulation. The hatched bands indicate statistical and systematic uncertainties for the **SM** background predictions. The normalisation of the signal processes is chosen arbitrarily for improved visualisation. The last bin of both histograms includes the overflow events.

CHAPTER 16

Signal Extraction with Boosted Decision Trees

A **MVA** is performed in **SR** to further separate the LFV signals from the backgrounds, and enhance the sensitivity of this analysis. More specifically, a dozen of discriminating variables, referred to as “features”, are selected and combined by a gradient-**BDT**, which is implemented using the XGBOOST package [27]. There are several reasons why **BDT** is chosen: (i) the goal of the **MVA** is to achieve maximum separation between signals and backgrounds using a small number of already well-separated kinematic variables, instead of exploring some complicated structures hidden in event topology, (ii) under such a goal, the potential performance gain from a more sophisticated algorithm like a **NN** is limited, (iii) a **BDT**-based algorithm is straightforward to implement and consumes only a moderate amount of computational resources, and (iv) the interpretability of a **BDT**-based algorithm is excellent.

The top production and decay signals are longer distinguished by the **BDT**. They are combined into a single signal class, just like all backgrounds are combined into a single background class. The training of the **BDT** depends entirely on **MC** samples that are statistically orthogonal to the samples used in the actual background estimation. More details on the configurations of the **BDT** are described in section 16.1. The input features are described in section 16.2. The output of the **BDT** is presented in section 16.3.

16.1 BDT Configuration

The LFV $e\mu$ mass of the top decay signal is bounded by the **SM** top quark mass, as is shown in Figure 14.1. On the contrary, the LFV $e\mu$ mass of the top production signal has no such restriction and often reaches TeV level. Therefore, a 150 GeV threshold is used to divide the **SR** into two **SRs** enriched in different signal modes. The **MVA** strategy is to combine two signal modes within each **SR** and train binary **BDTs** separately for each **SR**. In other words, only two signal datasets and two background datasets are needed.

Other aspects of the signal **MC** samples, such as the Lorentz structure and the flavor of the up-type quark involved in LFV interaction, are shown to have a relatively small impact on the kinematics of final state particles, as is shown in Figure 16.1. Therefore, they are not distinguished by the **BDT**. The construction of the signal datasets take a few steps. Firstly, all top production

signal samples, regardless of the Lorentz structure or the quark flavor, are normalised to the cross section of the vector-like operator with an $e\mu\tau\mu$ vertex, which is shown in Figure 12.2. Similarly, all top decay signals samples are normalised to the cross section of the vector-like operator. For each sample, a normalisation weight is calculated and is absorbed by the MC weight, which originally only covers per-event correction. These updated weights are eventually passed on to the BDT to weight each signal event. Secondly, all top production and decay signal samples are combined into a single dataset, which is then divided into two subsets using a 150 GeV threshold on LFV $e\mu$.

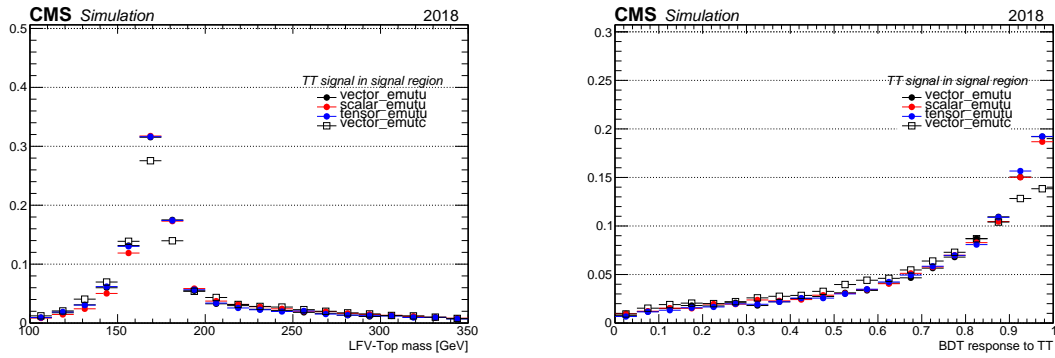


Figure 16.1: Normalized distribution in SR1. From left to right: LFV top mass, BDT shape

All background MC samples listed in Table 12.3 are used in the BDT training. The

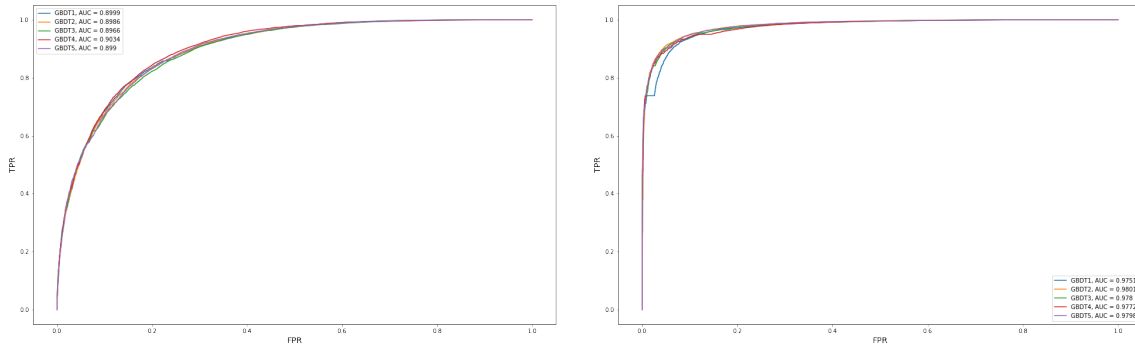


Figure 16.2: ROC curve with 5-fold cross validation. From left to right: BDT targeting TT (SR1), BDT targeting ST (SR2).

The MC dataset is divided into 5 subsets evenly. A 5-fold cross-validation scheme is deployed: the training set consists of 3 out of the 5 subsets, while validation/testing set each consists of 1 subset. The number of estimators is set to 1000, max depth 5, and sub-sample 0.8 with the evaluation metric set to "mlogloss". The results of the cross-validation study are shown in Figure 16.2. The trained BDT model targeting TT(ST) signal is used to evaluate each event in SR1(SR2): a BDT weight is assigned to each event, giving the likelihood that it originated from TT(ST) signal.

16.2 BDT Features

The observables used in training are referred to as "features" in this analysis. The following features are used for both ST and TT classifier:

- **MVA_ $M_{e\mu}$** : invariant mass of the Opposite-Sign $e\mu$ pair
- **MVA_ LFVePt**: p_T of the flavor-violating electron
- **MVA_ LFVmuPt**: p_T of the flavor-violating muon
- **MVA_ LFVTopmass**: invariant mass of the flavor-violating top-quark candidate
- **MVA_ Zmass**: invariant mass of Z boson candidate
- **MVA_ Jet2Btag**: b-tagging score of the jet with the second highest b-tagging score
- **MVA_ Mbl2**: invariant mass of the second b-jet+lepton system
- **MVA_ njet**: number of jets
- **MVA_ nbjet**: number of b-jets
- **MVA_ tM**: transverse mass of the W boson candidate (from standard model top-quark)
- **MVA_ lIDr**: ΔR between flavor-violating electron and muon
- **MVA_ SSee_ Zmass**: invariant mass of a Same-Sign di-electron pair
- **MVA_ Topmass**: invariant mass of the standard model top-quark candidate
- **MVA_ Met**: missing transverse energy

The following features are only used for TT classifier:

- **MVA_ Ht**: scalar sum of the p_T of all objects
- **MVA_ Mbl1**: invariant mass of the second bjet+lepton system
- **MVA_ JeDr**: ΔR between flavor-violating electron and a light jet (non-b-jet)
- **MVA_ JmuDr**: ΔR between flavor-violating muon and a light jet (non-b-jet)

The following features are used for ST classifier:

- **MVA_ BaPt**: p_T of the standard model lepton
- **MVA_ JetHt**: scalar sum of the p_T of jets

Distributions of selected features are shown in Figure 16.3-16.7.

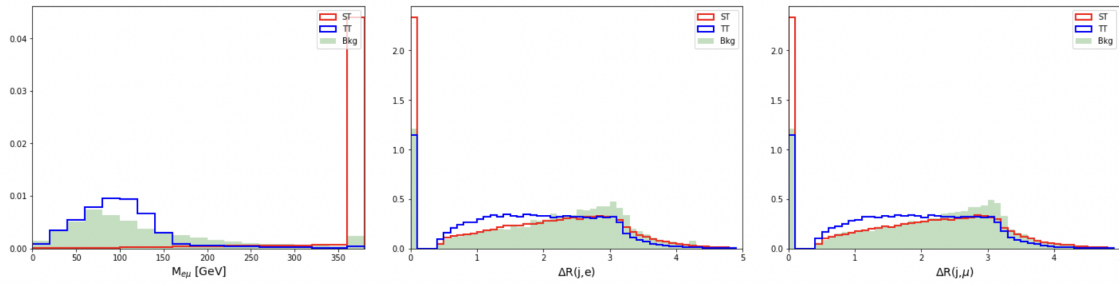


Figure 16.3: Normalized distribution of various features in SR. From to left to right: $M_{e\mu}$, M_{JeDr} , M_{JeDr} .

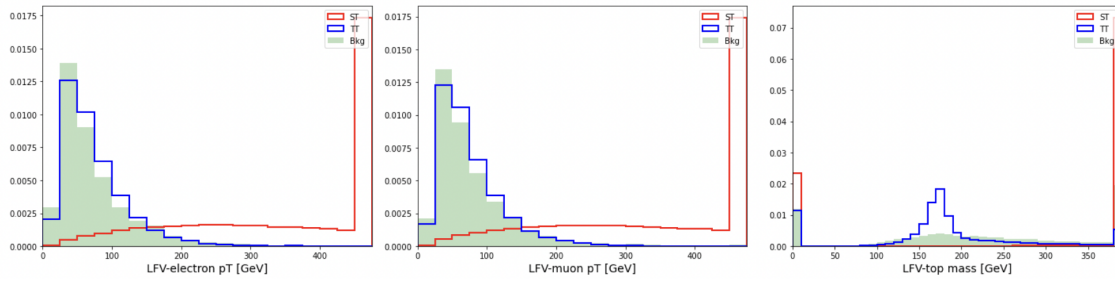


Figure 16.4: Normalized distribution of additional features in SR. From to left to right: M_{LFVePt} , M_{LFVmuPt} , $M_{\text{LFVTopmass}}$.

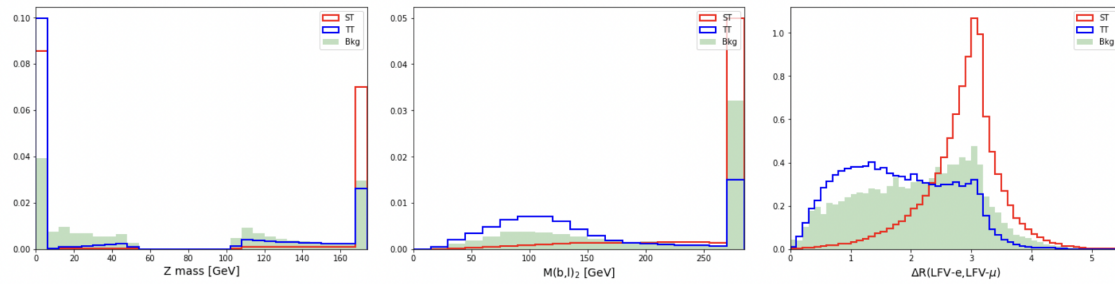


Figure 16.5: Normalized distribution of additional features in SR. From to left to right: M_{Zmass} , M_{Mbl2} , M_{IIDr} .

To determine the relative importance of these features, feature importance is extracted using the "gain" method and is shown in Figure 16.10.

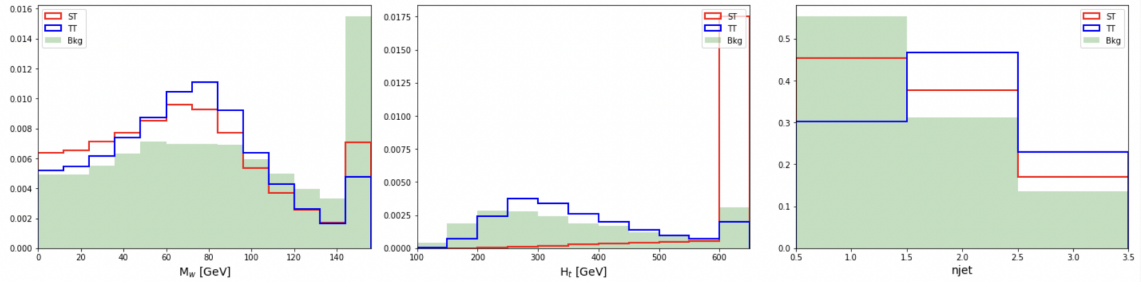


Figure 16.6: Normalized distribution of additional features in SR. From to left to right: MVA_tM, MVA_Ht, MVA_njet.

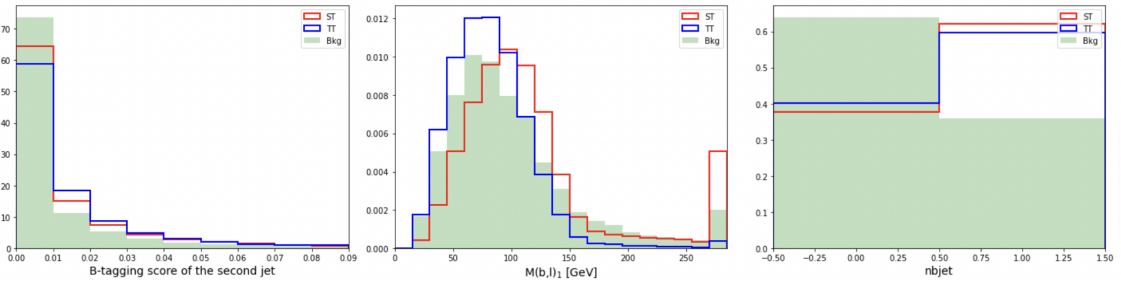


Figure 16.7: Normalized distribution of additional features in SR. From to left to right: MVA_Jet2Btag, MVA_Mbl1, MVA_nbjet.

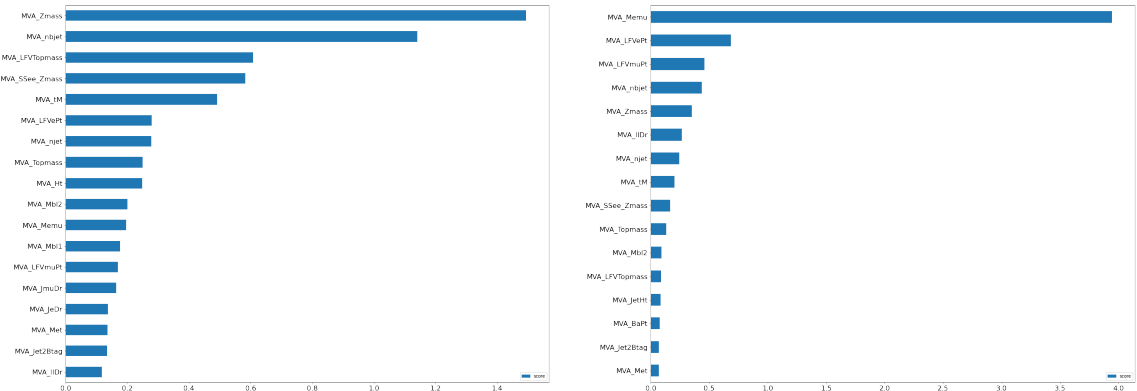


Figure 16.8: List of features ranked by their relative importance. From left to right: BDT targeting TT (SR1), BDT targeting ST (SR2)

16.3 BDT Output

The output of the BDTs are shown in Figure 16.11. Note: all but the nonprompt backgrounds are estimated with MC simulation. The nonprompt background backgrounds are estimated with

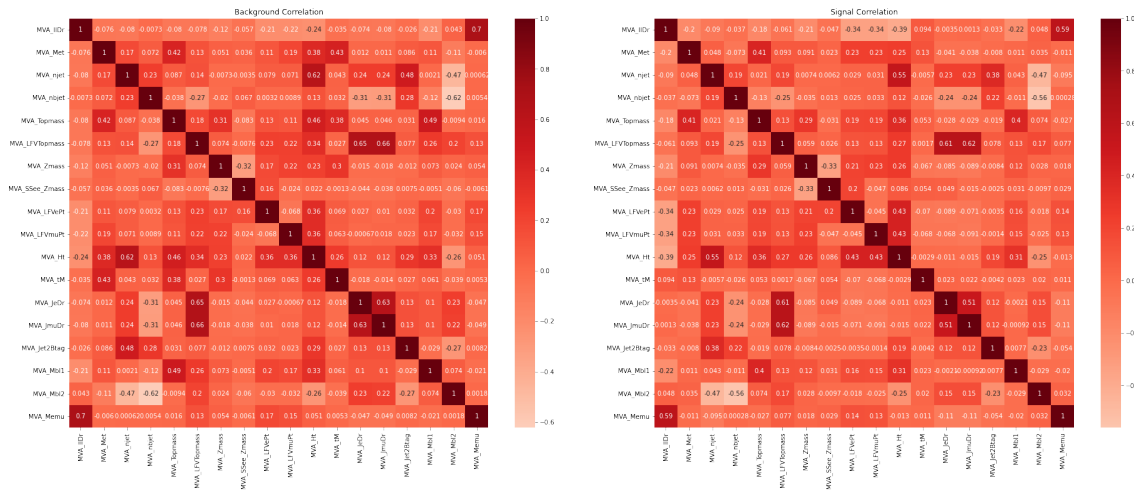


Figure 16.9: Correlation matrices (SR1), from left to right: background correlation, signal correlation.

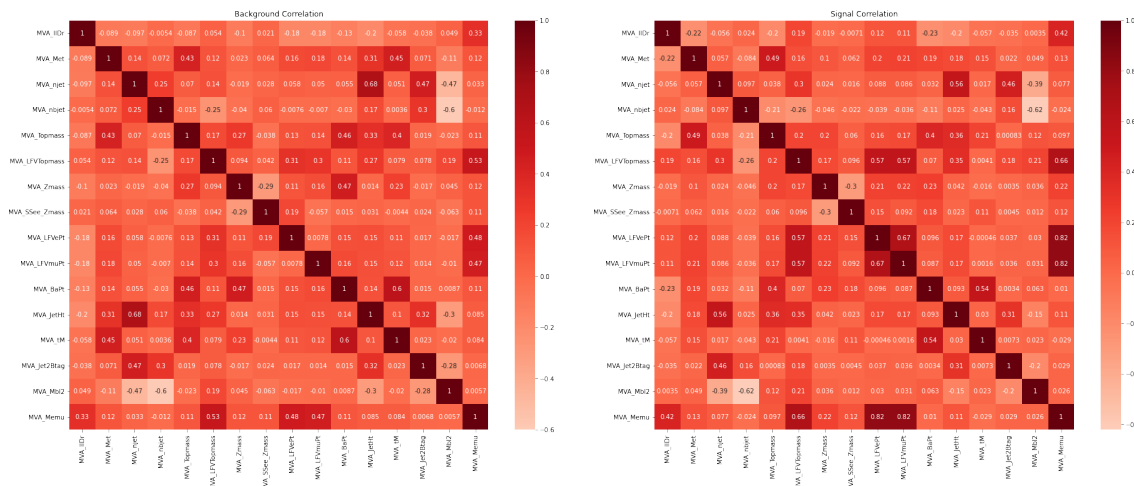


Figure 16.10: Correlation matrices (SR2), from left to right: background correlation, signal correlation.

the matrix method.

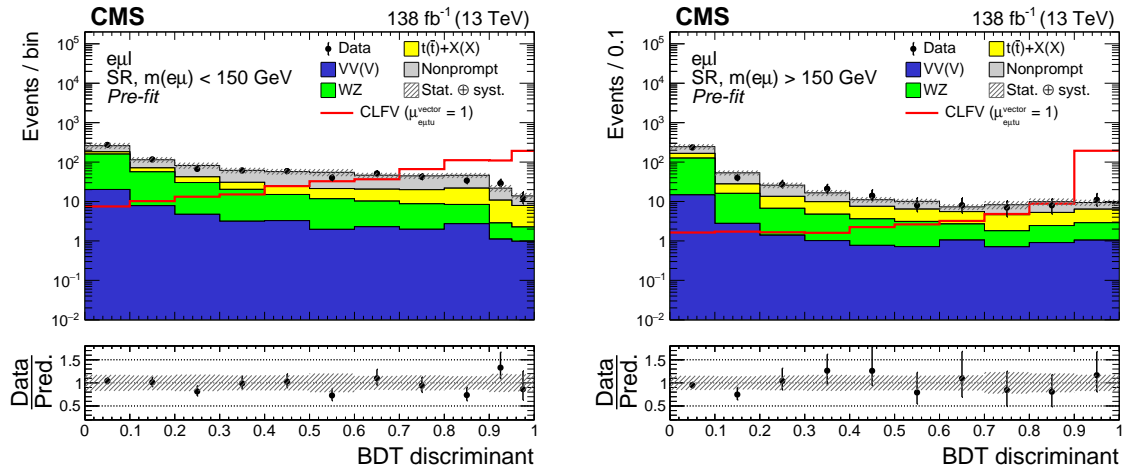


Figure 16.11: Distribution of BDT output, from left to right TT enriched SR (SR1), ST enriched SR (SR2)

CHAPTER 17

Systematic Uncertainties

For this analysis, we consider the following sources of systematic uncertainties:

17.1 Theoretical Uncertainties

- Prompt background cross-sections: These uncertainties arise from the uncertainties in the measured cross-sections of the processes which are used for scaling the MC to the luminosity of data. Systematic uncertainties associated to the prompt backgrounds WZ (as discussed in autorefsec:CRWZ), ZZ , ttW/ttZ/ttH, tZq are 6%, 6%, 15% and 20% respectively. Measurements performed on the WZ and ZZ processes currently have uncertainties of order 5-8% citeWZ2019 citeZZ2020. A 50 % normalization uncertainty is assigned to other smaller backgrounds.
- PDF: Using 100 the replicas of the NNPDF 3.1 set, the PDF uncertainties are obtained by taking the root mean square value of the variations. Only shape components of the variation is considered, this is done by normalizing LHE weights of the replicas such that the sum of the LHE wights (in SR) stays the same for each replica. This uncertainty treated as correlated. We consider this uncertainty for all the signals and major prompt backgrounds (i.e. WZ, TTZ, TTW and TTH).
- Renormalization and factorization scale: this uncertainty is evaluated by considering an envelope of 6 variation (μ_R 2, μ_R 0.5, μ_F 2, μ_F 0.5, μ_R and μ_F 2, μ_R and μ_F 0.5). This uncertainty is treated as correlated and only shape component is considered. We consider this uncertainty for all the signals and major prompt backgrounds (i.e. WZ, TTZ, TTW and TTH).
- ISR and FSR: initial state radiation and final state radiation uncertainty is evaluated by using the default set of variation (2 and 0.5). This uncertainty is treated as correlated and only shape component is considered. We consider this uncertainty for all the signals.

17.2 Nonprompt Uncertainties

We consider the following sources of systematic uncertainties (shape) associated with the *fake* efficiency.

- Prompt background correction
- Sample dependence
- Statistical uncertainty

An important systematic uncertainty comes from the estimate of prompt contamination in MR. As is discussed in [chapter 15](#), prompt backgrounds (estimated with MC) are subtracted from total event yields measured in MR (data). A flat 20 % uncertainty (α in Equation 17.1) is assigned to the event yields of the prompt background and the resulting variation of f is taken as the uncertainty, which is estimated to be between 5 % and 15 %.

$$f = \frac{n_{data}^T - (1 + \alpha)n_{Prompt\ MC}^T}{n_{data}^{T+\bar{T}} - (1 + \alpha)n_{Prompt\ MC}^{T+\bar{T}}}. \quad (17.1)$$

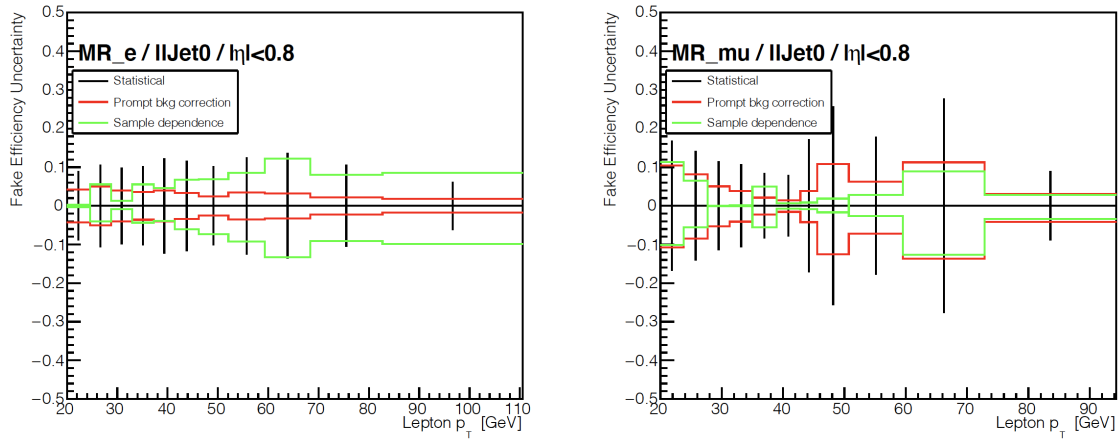


Figure 17.1: Comparison of different components of the uncertainties associated to the *fake* efficiency measured in 2017 dataset ($n_{jet}=0$ bin, $|\eta| < 0.8$ bin). From left to right: electron f uncertainty, muon f uncertainty. Note: black uncertainty bars represent statistical uncertainties.

Sample dependence is one of the primary sources of systematic uncertainty and it is concerned with the possibility that the *fake* efficiency f measured in MR may not reflect characteristics of the non-prompt leptons in SR due to different compositions of backgrounds. This type of uncertainty is estimated by introducing a variation factor β between the proportions of Same-Flavor and Opposite-Flavor pairs in MR. For example, electron f can be calculated as (prompt background correction is ignored from the equation):

$$f_e = \frac{(1 + \beta)n_{F,e+e}^T + (1 - \beta)n_{F,e+\mu}^T}{(1 + \beta)n_{F,e+e}^{T+\bar{T}} + (1 - \beta)n_{F,e+\mu}^{T+\bar{T}}}. \quad (17.2)$$

A 20% variation (β) is assigned the resulting variation of f is taken as the uncertainty, which is estimated to be between 3 % and 15 %.

A comparison of different sources of systematic uncertainties (using 2017 datasets) are shown in Figure 17.1 and Figure 17.2. A complete summary of uncertainties associated to *real* and *fake* efficiency can be found in.

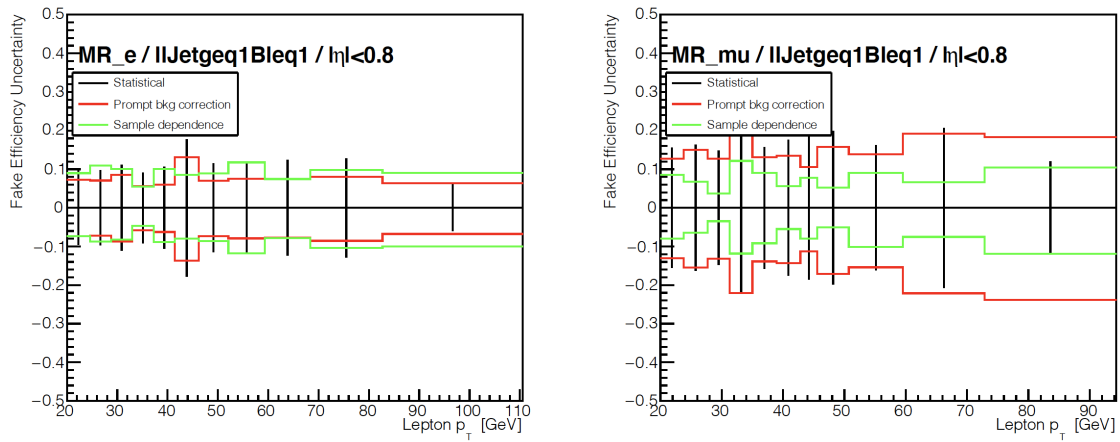


Figure 17.2: Comparison of different components of the uncertainties associated to the *fake* efficiency measured in 2017 dataset ($n_{\text{jets}} > 0$ bin, $|\eta| < 0.8$ bin). From left to right: electron f uncertainty, muon f uncertainty. Note: black uncertainty bars represent statistical uncertainties.

In addition to statistical uncertainties, MC uncertainties are propagated to the *real* efficiency as sources of systematic uncertainties:

- Statistical uncertainty
- MC uncertainty
 - Electron reconstruction
 - b-tagging
 - JES/JER
 - Pile-up reweighting
 - ECAL prefire

The uncertainties associated to the *real* efficiency are relatively small when compared to the *fake* efficiency uncertainties. A comparison of different sources of *real* efficiency uncertainties are shown in Figure 17.3.

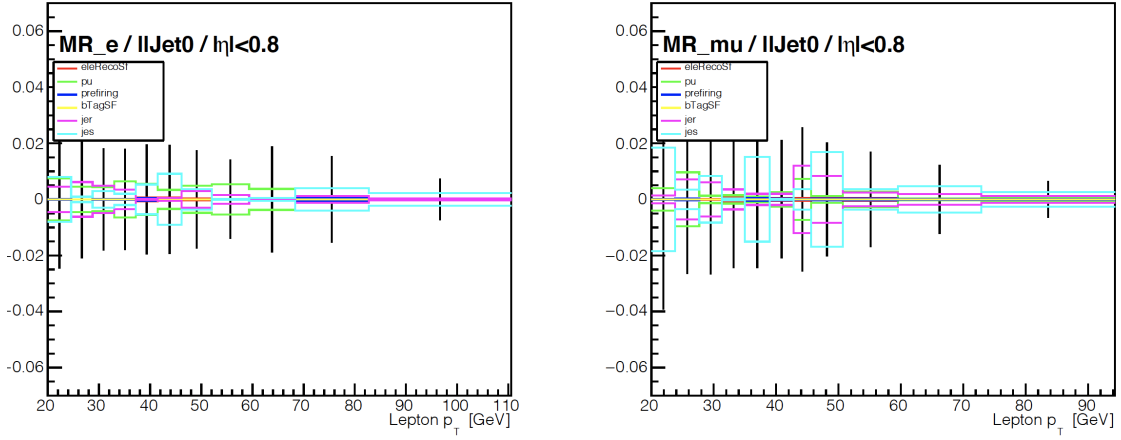


Figure 17.3: Comparison of different components of the uncertainties associated to the *real* efficiency measured in 2017 dataset ($n_{\text{jets}}=0$ bin, $|\eta| < 0.8$ bin). From left to right: electron r uncertainty, muon r uncertainty. Note: black uncertainty bars represent statistical uncertainties.

In addition to the four components (i.e. r_e , r_μ , f_e , f_μ) of the shape uncertainties associated to the measurement of r and f , a fifth component is considered that accounts for the potential bias introduced when implementing 3D matrix method: Four out of the eight regions that appear on the lefthand side of the Equation 15.6 (i.e. $N^{\bar{T}TT}$, $N^{\bar{T}T\bar{T}}$, $N^{T\bar{T}T}$, $N^{T\bar{T}\bar{T}}$) are selected by requiring the leading lepton to fail the *tight* criteria (See Table 15.2). Effectively this means that the isolation requirement is reversed for leading lepton that enter these four regions. Selecting the leading lepton by a loose/reversed isolation requirement is not ideal since the leading lepton is required to match with iso-trigger. To account for this bias, a 50 % uncertainty is assigned to the f_1 (*fake* efficiency of the leading lepton) for events that enter these four regions. The variation of the non-prompt estimate due to trigger matching is largely covered by this uncertainty (See Figure 17.4).

The five components of the uncertainties discussed in this section are propagated through the matrix inversion. The resulting variations of the non-prompt estimates are taken as the shape uncertainties assigned to non-prompt backgrounds. These uncertainties are treated as correlated between the years. In addition to these shape uncertainties, an overall normalization uncertainty of 10% is assigned in order to cover any other potential bias.

17.3 Diboson Uncertainties

Mismodeling of the jet multiplicity is observed in WZ control region (See autorefsec:CRWZ). This is largely due to the fact that jets generated from diboson samples are coming from radiation (due to our triplepton selection). Therefore, the accuracy of the jet multiplicity distribution is suboptimal. To take this into account, a dedicated jet-dependent uncertainty is assigned to each

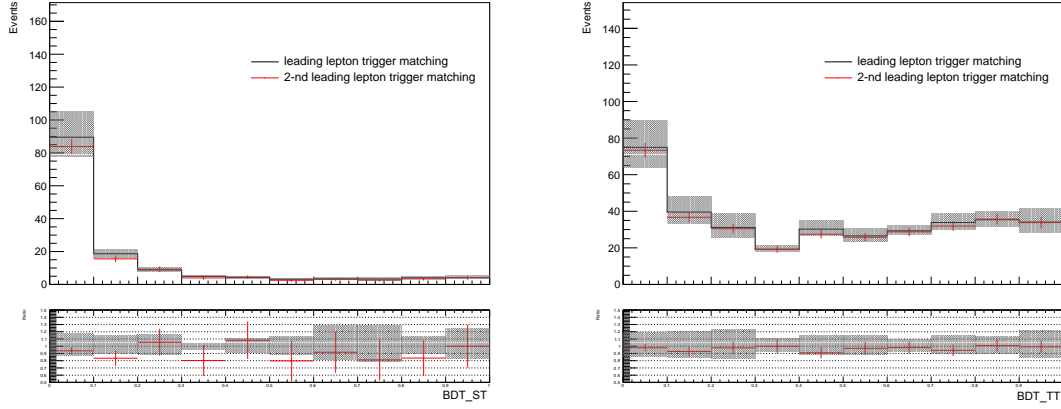


Figure 17.4: The impact of matching leptons to trigger objects on non-prompt estimate. From left to right: non-prompt estimate in ST enriched SR, non-prompt estimate in TT enriched SR. The nominal configuration of the matrix method is to match the leading lepton with trigger objects. Matching the 2nd-leading with the trigger objects is taken as an alternative to evaluate the robustness of the non-prompt estimate. The uncertainty band only covers the variation of the non-prompt estimate as a result of varying leading lepton f by 50 %. Uncertainty bars only include statistical uncertainties.

event. This uncertainty is determined using diboson control region:

- Exactly three leptons (any flavor composite),
- OnZ,
- $\text{MET} > 85\text{GeV}$,
- Veto events with at least one b-tag jet ($p_T > 20\text{GeV}$).

The jet multiplicity distributions are shown in Figure 17.5. For each year, a scale factor parameterized as bins of jet multiplicity is derived:

$$\epsilon = \frac{N_{data} - N_{VVV} - N_{TX} - N_{TTbar} - N_{others}}{N_{VV}}. \quad (17.3)$$

The scale factor ϵ is used to estimate the uncertainty, denoted by Δ :

$$\Delta = |1 - \epsilon| \quad (17.4)$$

This uncertainty is about 10-20%, as is shown in Figure 17.6.

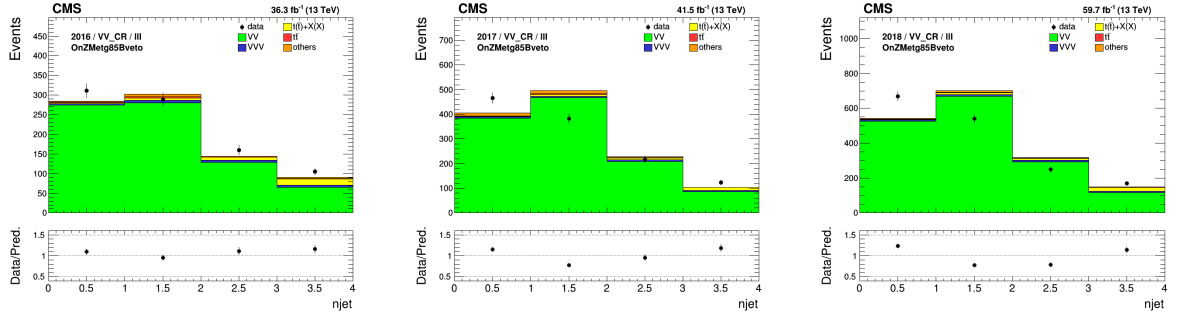


Figure 17.5: Diboson control region, from left to right: 2016, 2017 and 2018 datasets.

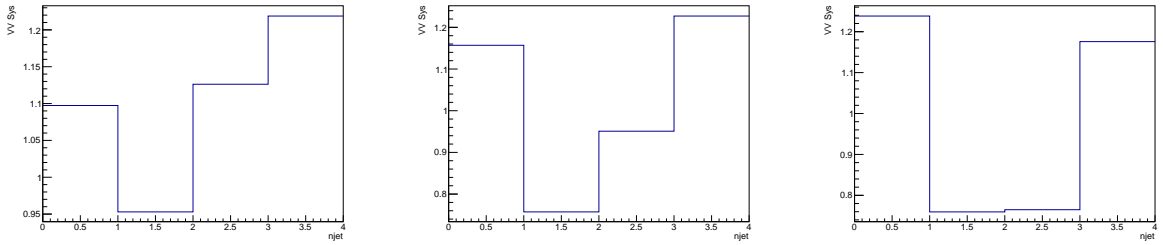


Figure 17.6: Scale factors derived from diboson control region, from left to right: 2016, 2017 and 2018 datasets.

17.4 Other Experimental Uncertainties

- Lepton reconstruction, identification, and isolation: electron/muon reconstruction uncertainties are provided by relevant POG. Uncertainties associated with lepton identification and isolation scale factors were evaluated by the authors of the mvaTOP ID citemvaTOP. The statistical components of these uncertainties are treated as uncorrelated while the other components are treated as fully correlated. For high p_T muons ($p_T > 200\text{GeV}$), an additional uncertainty (denoted "mulDHighPt") is assigned and it increases linearly from 0 to 10% (200GeV-1000GeV) and is capped at 10% after 1000GeV.
- Muon scale uncertainties: uncertainties on muon momentum (denoted "MuonScale") are assigned using different method according to the muon p_T : Rochester algorithm is used for low p_T muons ($p_T < 200\text{GeV}$), while GECiteGEmethod method is used for high p_T muons ($p_T > 200\text{GeV}$).
- b-tagging: b-tagging efficiency and uncertainty is provided by the BTV group.
- trigger scale factor: trigger scale factors are set to 1 and a flat 2% uncertainty is assigned. This uncertainty is treated as uncorrelated.

- **Luminosity:** The uncertainties of 1.0%, 2.0% and 1.5% are assigned to the integrated luminosity for 2016, 2017, 2018, respectively. When processing the run 2 data the individual year uncertainties are treated as uncorrelated while there are two additional correlated luminosity uncertainties; firstly, 0.6%, 0.9% and 2.0% for each year respectively to account for the correlation between 2016, 2017 and 2018 data and secondly, 0.6%, 0.2% for 2017 and 2018 respectively to account for the correlation between 2017 and 2018 data.
- **jet energy scale and resolution:** The jet energy scale, JES, and jet energy resolution, JER, are centrally provided by the JET/MET POG [citeJEC](#). Variations of jet energy scale and resolution are propagated to the MET and b-tagging SF.
- **pile-up reweighting:** The measured minimum bias cross-section is varied up and down by 4.6%. This uncertainty is treated as correlated.
- **MET unclustered missing energy** is considered and treated as uncorrelated across the years.
- **L1 ECAL prefiring:** In the 2016 and 2017 datasets, L1 EGamma triggers fired early causing many uninteresting events to be recorded while the later interesting events were rejected. Since this effect is not present in the MC simulation, a tool [citeECALPre](#) provided by the L1 DPG is used to reweight MC events. This uncertainty is treated as correlated.
- **HEM15/16 Issue:** The HEM15/16 issue refers to two HCAL modules whose power supply died in the middle of the data taking (runs ≥ 319077 , i.e. last certified run of 2018B, and all of 2018C+D). The HEM issue is likely a very small effect but we still have to check it following the procedure in [citeHEM](#)

Uncertainties related to B-tagging scale factors are split into different sources. For b and udsg jets, we applied lf, hf, hfstats1/2, and lfstats1/2 uncertainties. For c jets, we applied cferr1/2 uncertainties. Correlations between different sources are specified in the [Table 17.1](#).

Table 17.1: A hyphen (–) denotes that a source is not correlated between the different years.

Source	Correlated	Description
lf	✓	udsg+c jets in heavy flavor region
hf	✓	b+c jets in light flavor region
hfstats1	–	Linear fluctuations of c jets
hfstats2	–	Quadratic fluctuations of c jets
lfstats1	–	Linear fluctuations of udsg jets
lfstats2	–	Quadratic fluctuations of udsg jets
cferr1	✓	Linear fluctuations of c jets
cferr2	✓	Quadratic fluctuations of c jets

17.4.1 Jet energy scale uncertainties

Uncertainties associated with JES are split into their 27 components and properly treated the correlations of the split uncertainties by year as recommended by the JETMET POG and described in the Table 17.2.

Table 17.2: Summary of the sources of uncertainty related to the JECs. A hyphen (–) denotes that a source is not correlated between the different years.

Source	Correlated	Source	Correlated
AbsoluteStat	-	RelativePtHF	✓
AbsoluteScale	✓	RelativeBal	✓
AbsoluteMPFBias	✓	RelativeSample	-
Fragmentation	✓	RelativeFSR	-
SinglePionECAL	✓	RelativeStatFSR	✓
SinglePionHCAL	✓	RelativeStatEC	-
FlavorQCD	✓	RelativeStatHF	-
TimePtEta	-	PileUpDataMC	✓
RelativeJEREC1	-	PileUpPtRef	✓
RelativeJEREC2	-	PileUpPtBB	✓
RelativeJERHF	✓	PileUpPtEC1	✓
RelativePtBB	✓	PileUpPtEC2	✓
RelativePtEC1	-	PileUpPtHF	✓
RelativePtEC2	-		

B-tagging scale factors and MET vector are computed for each of the JES templates and treated as uncertainties that are fully correlated to the respective JES sources.

The overall systematic uncertainty on background is estimated to be about 15 % in SR.

A comparison of different sources of systematic uncertainties of the background estimates in SR is shown in Figure 17.7.

A comparison of different sources of systematic uncertainties of the signal estimates in SR is shown in Figure 17.8.

Representative range of systematic uncertainties on background and signal MC samples are summarized in citeSysError. These uncertainties are extracted from the last bins of the BDT distribution.

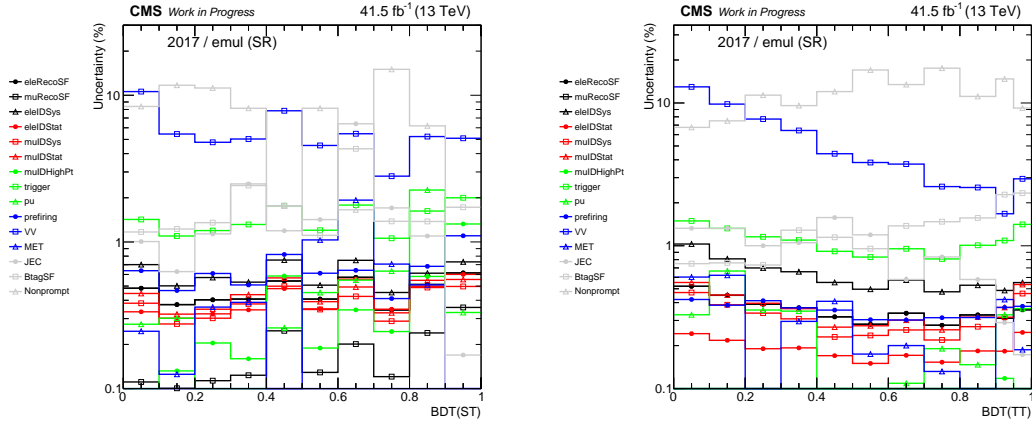


Figure 17.7: Distributions of relative uncertainties on total expected backgrounds as a function of BDT output in ST enriched SR (left), TT enriched SR (right). Luminosity and normalization uncertainties are not included in these plots. JES, JER and HEM are combined into "JEC". Sources of b-tagging uncertainties listed in Table 17.1 are combined into "BtagSF".

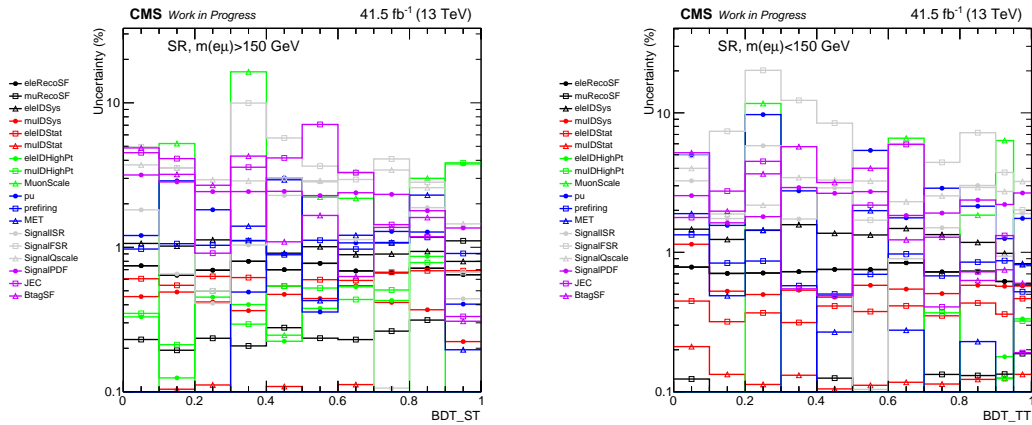


Figure 17.8: Distributions of relative uncertainties on signal ($C_{e\mu\tau}^{vector}$ is used as an example) as a function of BDT output in ST enriched SR (left), TT enriched SR (right). Luminosity and normalization uncertainties are not included in these plots. JES, JER and HEM are combined into "JEC". Sources of b-tagging uncertainties listed in Table 17.1 are combined into "BtagSF".

CHAPTER 18

Statistical Analysis

18.1 Profile Likelihood Fit

A binned likelihood function $\mathcal{L}(\mu, \theta)$ is constructed to perform the statistical analysis on the BDT discriminator distributions. The parameter of interest (POI), denoted by μ , is the signal strength that governs the cross section of the ST and TT signals simultaneously. All the uncertainties are incorporated into the likelihood function as nuisance parameters, denoted by θ . The uncertainties that affect the shape of the BDT discriminator distributions are considered with Gaussian distributions while other uncertainties that only affect the normalizations are considered with log-normal distributions.

To control the systematic uncertainties, a profile likelihood fit is performed simultaneously in six regions (three data-taking years and two signal regions) by maximizing the likelihood function $\mathcal{L}(\hat{\mu}, \hat{\theta})$. Both $\hat{\mu}$ and $\hat{\theta}$ are the maximum likelihood estimators for the signal strength and nuisance parameters, respectively. The scaling of the signal for the purpose of a better visualization is optimized based on "VecU" signal, and it is kept the same for the other signals to compare their cross sections relative to each other. The post-fit distributions of the BDT discriminator are shown in Fig. 18.1-???. The most prominent uncertainties affecting the likelihood fit are the statistical uncertainties on MC samples.

18.2 Upper Limits

The results for the one-dimensional limits are summarized in Table 18.1. Assuming a linear relationship between $\mathcal{B}(t \rightarrow e\mu c)$ and $\mathcal{B}(t \rightarrow e\mu c)$ in the case of nonvanishing signals, the two-dimensional limits can also be obtained through interpolation (see Fig. 18.2). This analysis constitutes the most stringent limits on these processes to date.

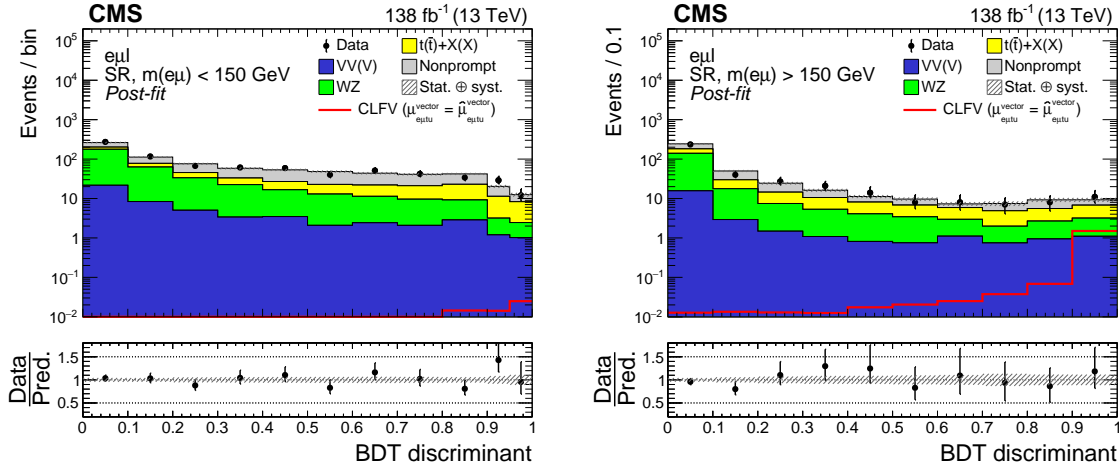


Figure 18.1: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with vector-like operator involving an Up quark. The three data-taking years are aggregated for better visualization.

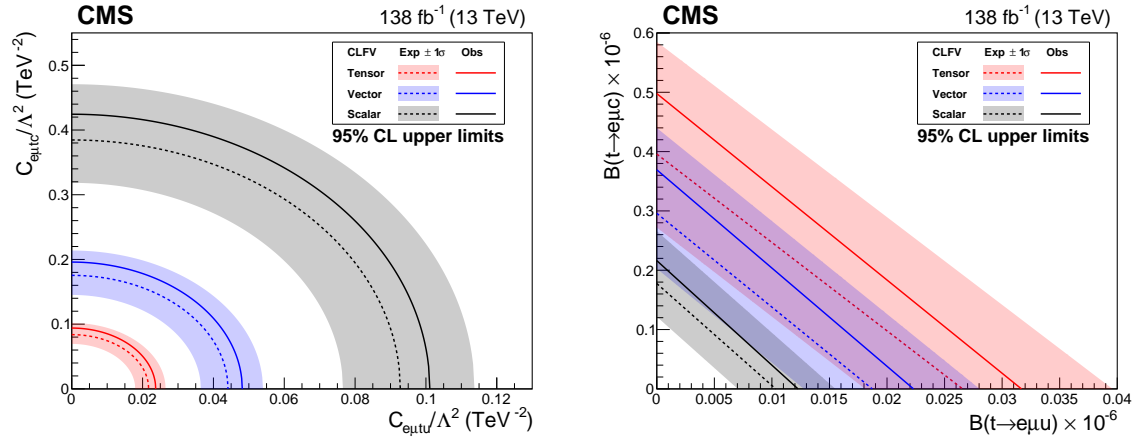


Figure 18.2: Two-dimensional upper limits on the Wilson coefficients (left) and the branching ratios (right).

CLFV coupling	Lorentz structure	$C_{e\mu tq}/\Lambda^2$ (TeV ⁻²)		$\mathcal{B}(t \rightarrow e\mu q) \times 10^{-6}$	
		exp (68% range)	obs	exp (68% range)	obs
$e\mu tu$	tensor	0.022 (0.018–0.026)	0.024	0.027 (0.018–0.040)	0.032
	vector	0.044 (0.036–0.054)	0.048	0.019 (0.013–0.028)	0.022
	scalar	0.093 (0.077–0.114)	0.101	0.010 (0.007–0.016)	0.012
$e\mu tc$	tensor	0.084 (0.069–0.102)	0.094	0.396 (0.272–0.585)	0.498
	vector	0.175 (0.145–0.214)	0.196	0.296 (0.203–0.440)	0.369
	scalar	0.385 (0.318–0.471)	0.424	0.178 (0.122–0.266)	0.216

Table 18.1: Upper limit on the LFV signal using the full Run 2 data set.

Part IV

Outlook for CLFV Searches Using Top Quarks

CHAPTER 19

Inclusive Signal Generation

19.1 UFO Model

19.2 EFT Reweighting

CHAPTER 20

Object and Event Selection

20.1 Object Selection

20.2 Event Selection

CHAPTER 21

Expected Sensitivity

21.1 Asimov Fit

21.2 Expected Upper Limits

CHAPTER 22

Run-2 Combination

22.1 Correlation Scheme

22.2 Minimal Flavor Violation Interpretation

CHAPTER 23

Summary and Conclusions

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APPENDIX A

List of Trigger Paths

Table A.1: Triggers used to record events during data taking in 2016.

Dataset	Trigger Path
SingleMuon	HLT_IsoMu22_eta2p1, HLT_IsoTkMu22_eta2p1
	HLT_IsoMu24, HLT_IsoTkMu24
	HLT_Mu50, HLT_TkMu50, HLT_Mu45_eta2p1
SingleElectron	HLT_Ele25_eta2p1_WPTight_Gsf
	HLT_Ele27_WPTight_Gsf
	HLT_Ele27_eta2p1_WPTight_Gsf
	HLT_Ele32_eta2p1_WPTight_Gsf
	HLT_Ele105_CaloldVT_GsfTrkIdT
	HLT_Ele115_CaloldVT_GsfTrkIdT
DoubleMuon	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL
	HLT_Mu17_TrkIsoVVL_TkMu8_TrkIsoVVL
	HLT_TkMu17_TrkIsoVVL_TkMu8_TrkIsoVVL
	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ
	HLT_Mu17_TrkIsoVVL_TkMu8_TrkIsoVVL_DZ
	HLT_TkMu17_TrkIsoVVL_TkMu8_TrkIsoVVL_DZ
	HLT_Mu30_TkMu11, HLT_TripleMu_12_10_5
DoubleEG	HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL
	HLT_DoubleEle33_CaloldL_MW
	HLT_DoubleEle33_CaloldL_GsfTrkIdVL
	HLT_DoubleEle33_CaloldL_GsfTrkIdVL_MW
MuonEG	HLT_Mu23_TrkIsoVVL_Ele8_CaloldL_TrackIdL_IsoVL
	HLT_Mu23_TrkIsoVVL_Ele8_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu8_TrkIsoVVL_Ele23_CaloldL_TrackIdL_IsoVL
	HLT_Mu30_Ele30_CaloldL_GsfTrkIdVL
	HLT_Mu33_Ele33_CaloldL_GsfTrkIdVL
	HLT_DiMu9_Ele9_CaloldL_TrackIdL
	HLT_Mu8_DiEle12_CaloldL_TrackIdL

Table A.2: Triggers used to record events during data taking in 2017.

Dataset	Trigger Path
SingleMuon	HLT_IsoMu24_eta2p1
	HLT_IsoMu27
	HLT_Mu50
	HLT_OldMu100
	HLT_TkMu100
SingleElectron	HLT_Ele32_WPTight_Gsf_L1DoubleEG
	HLT_Ele35_WPTight_Gsf
	HLT_Ele115_CaloldVT_GsfTrkIdT
DoubleMuon	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL
	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ
	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8
	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8
	HLT_Mu19_TrkIsoVVL_Mu9_TrkIsoVVL_DZ_Mass3p8
	HLT_Mu37_TkMu27
	HLT_TripleMu_12_10_5
	HLT_TripleMu_10_5_5_DZ
	HLT_TripleMu_5_3_3_Mass3p8to60_DZ
DoubleEG	HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL
	HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL
	HLT_DoubleEle25_CaloldL_MW
	HLT_DoubleEle33_CaloldL_MW
	HLT_DiEle27_WPTightCaloOnly_L1DoubleEG
MuonEG	HLT_Mu23_TrkIsoVVL_Ele12_CaloldL_TrackIdL_IsoVL
	HLT_Mu23_TrkIsoVVL_Ele12_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu8_TrkIsoVVL_Ele23_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu12_TrkIsoVVL_Ele23_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu27_Ele37_CaloldL_MV
	HLT_Mu37_Ele27_CaloldL_MV
	HLT_DiMu9_Ele9_CaloldL_TrackIdL
	HLT_DiMu9_Ele9_CaloldL_TrackIdL_DZ
	HLT_Mu8_DiEle12_CaloldL_TrackIdL

Table A.3: Triggers used to record events during data taking in 2018.

Dataset	Trigger Path
SingleMuon	HLT_IsoMu24
	HLT_IsoMu27
	HLT_Mu50
	HLT_OldMu100
	HLT_TkMu100
EGamma	HLT_Ele32_WPTight_Gsf
	HLT_Ele115_CaloldVT_GsfTrkIdT
	HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL
	HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL
	HLT_DoubleEle25_CaloldL_MW
	HLT_DiEle27_WPTightCaloOnly_L1DoubleEG
DoubleMuon	HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8
	HLT_Mu37_TkMu27
	HLT_TripleMu_12_10_5
	HLT_TripleMu_10_5_5_DZ
	HLT_TripleMu_5_3_3_Mass3p8to60_DZ
MuonEG	HLT_Mu23_TrkIsoVVL_Ele12_CaloldL_TrackIdL_IsoVL
	HLT_Mu23_TrkIsoVVL_Ele12_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu8_TrkIsoVVL_Ele23_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu12_TrkIsoVVL_Ele23_CaloldL_TrackIdL_IsoVL_DZ
	HLT_Mu27_Ele37_CaloldL_MV
	HLT_Mu37_Ele27_CaloldL_MV
	HLT_DiMu9_Ele9_CaloldL_TrackIdL_DZ
	HLT_Mu8_DiEle12_CaloldL_TrackIdL

APPENDIX B

Signal Region Distributions with MC simulation

Distributions of various variables in **SR**, which are used in the **BDT** training. More details on these input features are described in [section 16.2](#). The data are shown as filled points and the **SM** background predictions as histograms. The VV(V) background includes ZZ and triboson production, while the $t\bar{t} + X(X)$ component includes $t\bar{t}W$, $t\bar{t}Z$, $t\bar{t}H$, tZq , and smaller backgrounds containing one or two top quarks plus a boson or quark. All backgrounds are estimated using **MC** simulation. The hatched bands indicate statistical and systematic uncertainties for the **SM** background predictions. The normalisation of the signal processes is chosen arbitrarily for improved visualisation.

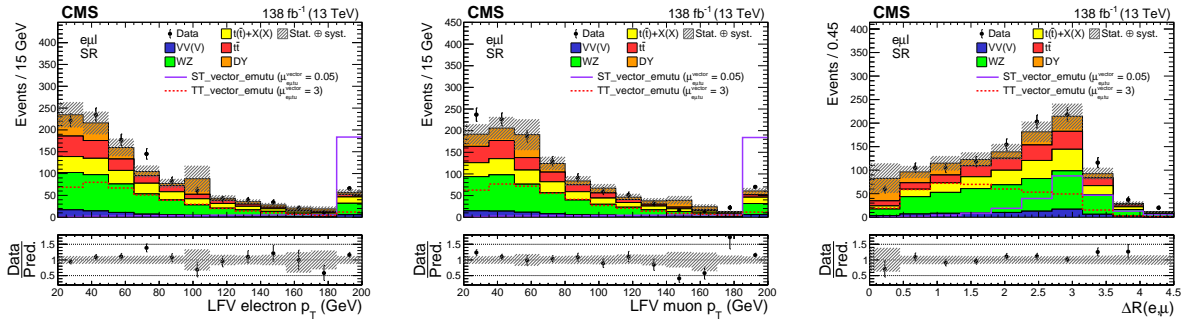


Figure B.1: Distributions of LFV electron p_T (left), LFV muon p_T (middle), and the opening angle between LFV electron and LFV muon (right).

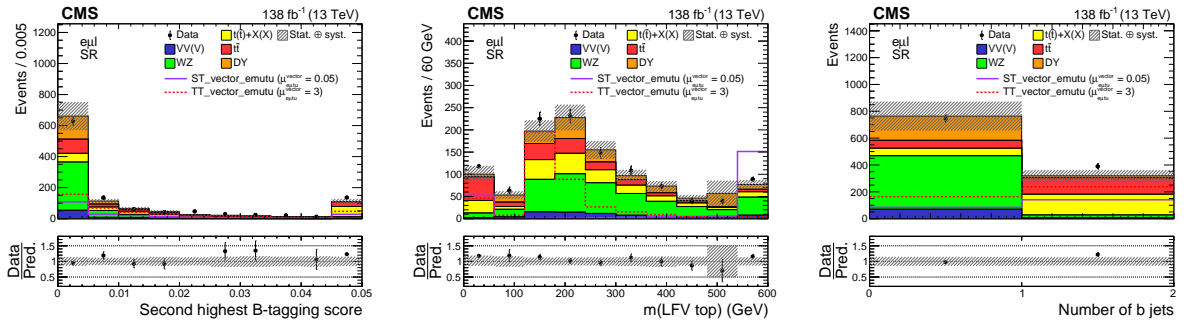
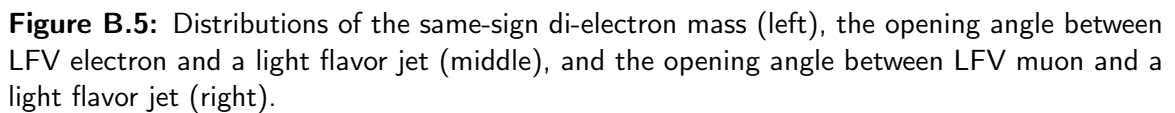
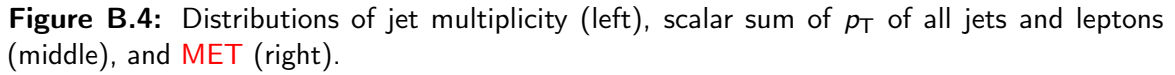
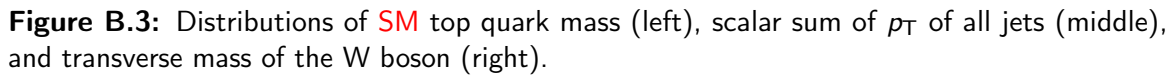


Figure B.2: Distributions of the second highest **DEEPIET** score (left), LFV top mass (middle), b jet multiplicity (right).



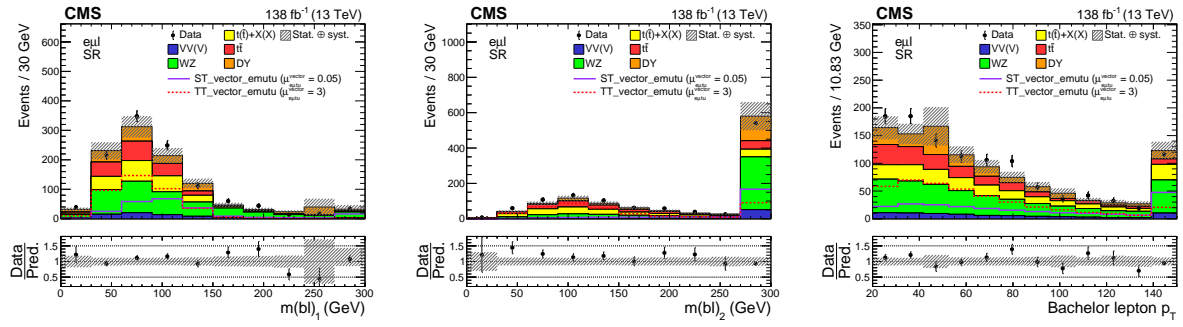


Figure B.6: Distributions of the mass of the first m_{bl} system (left), the mass of the second m_{bl} system (middle), and standalone lepton p_T (right).

APPENDIX C

Nuisance Parameter Impact

The observed and expected ($\mu_{exp} = 0$) impacts of the nuisance parameters on the profile likelihood fit are shown in Figure C.1-C.2. They were computed using the following commands and plotted below for full run 2.

The expected ($\mu_{exp} = 1$) impacts of the nuisance parameters on the profile likelihood fit are shown in Figure C.3.

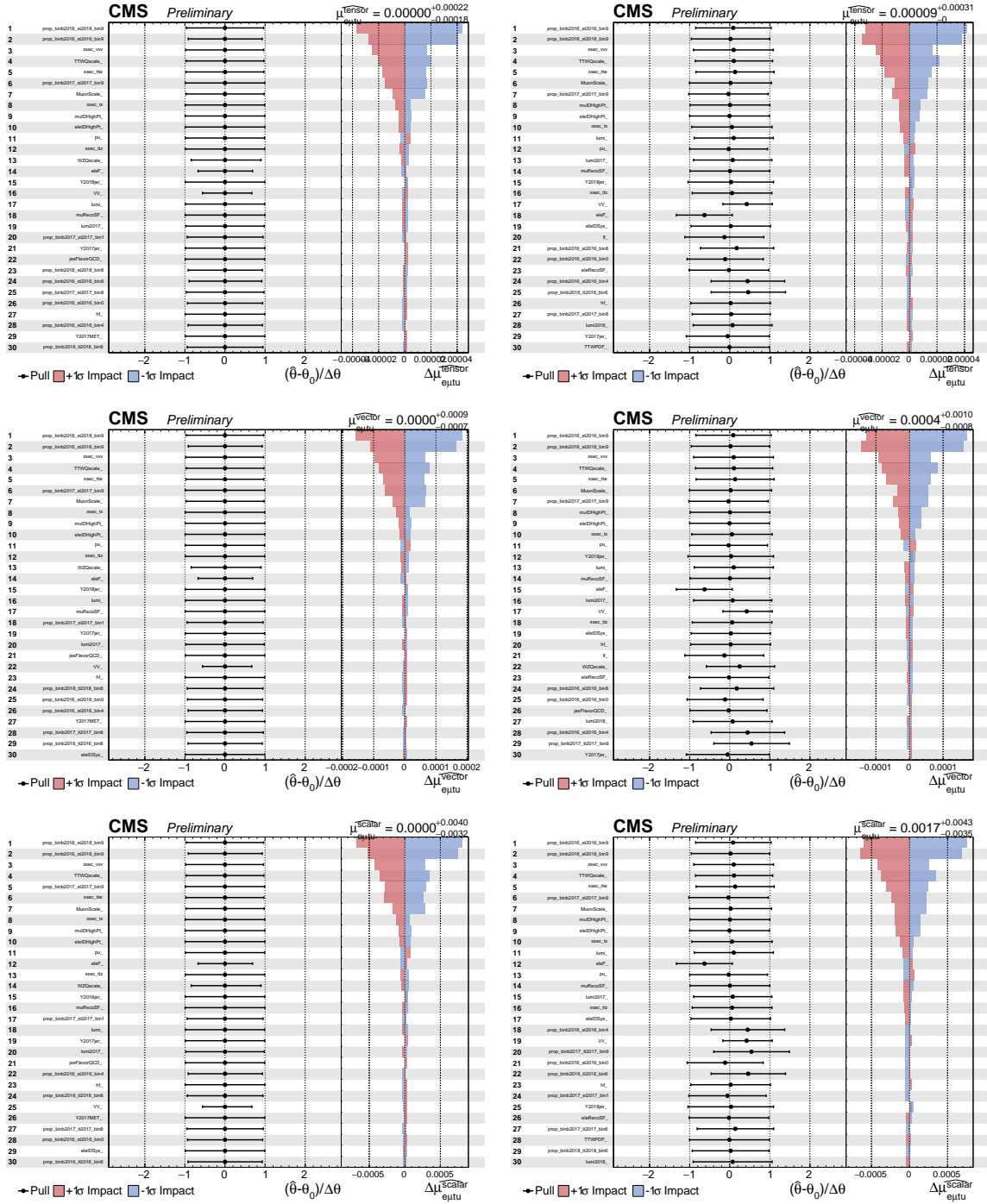


Figure C.1: Impacts of nuisance parameters for run II limit setting. From top to bottom: $e_{\mu\tau u}$ -tensor, $e_{\mu\tau u}$ -vector, $e_{\mu\tau u}$ -scalar. From left to right: expected impact (expected signal strength at 0), observed impact.

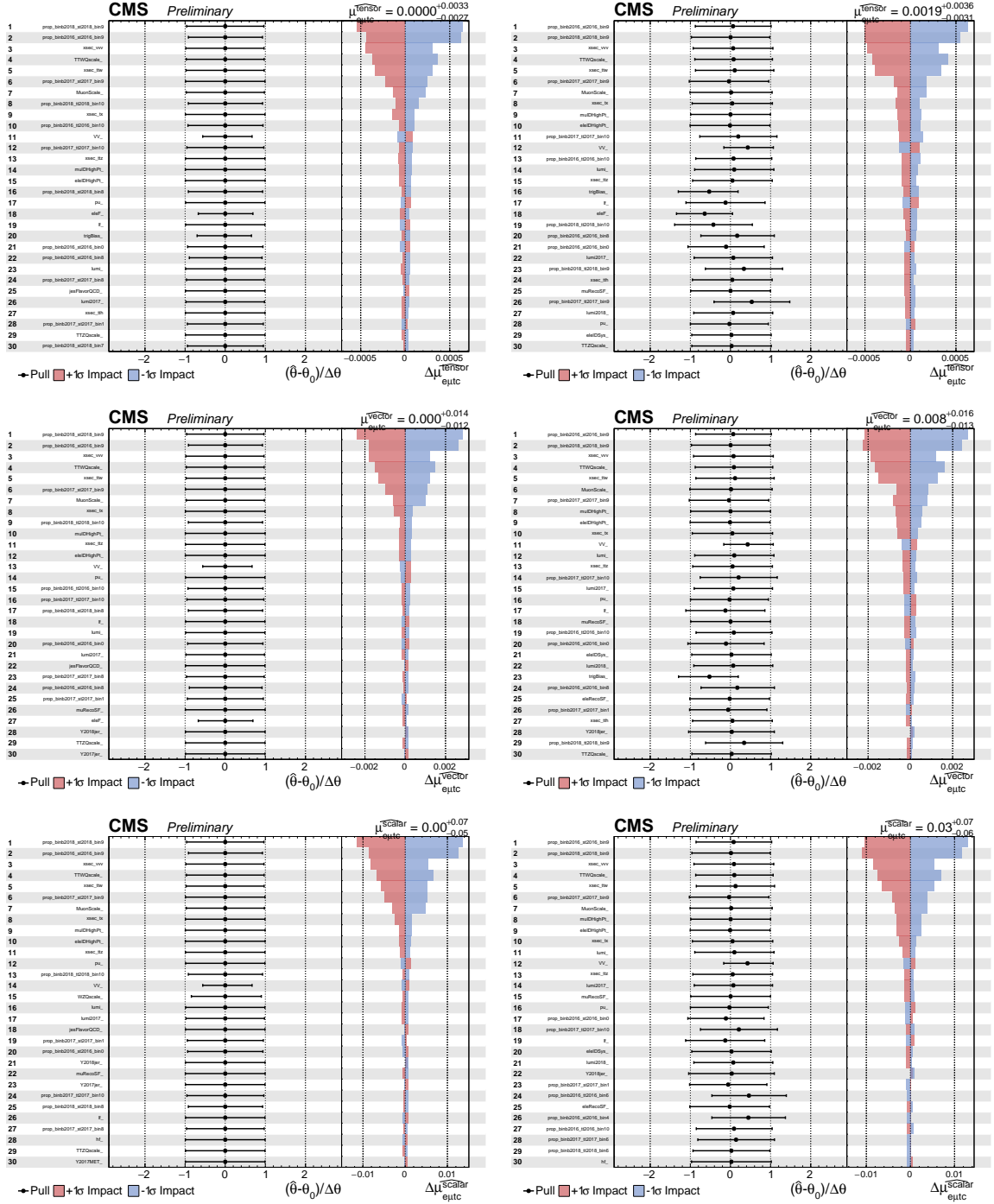


Figure C.2: Impacts of nuisance parameters for run II limit setting. From top to bottom: $e\mu_{tc}$ -tensor, $e\mu_{tc}$ -vector, $e\mu_{tc}$ -scalar. From left to right: expected impact (expected signal strength at 0), observed impact.

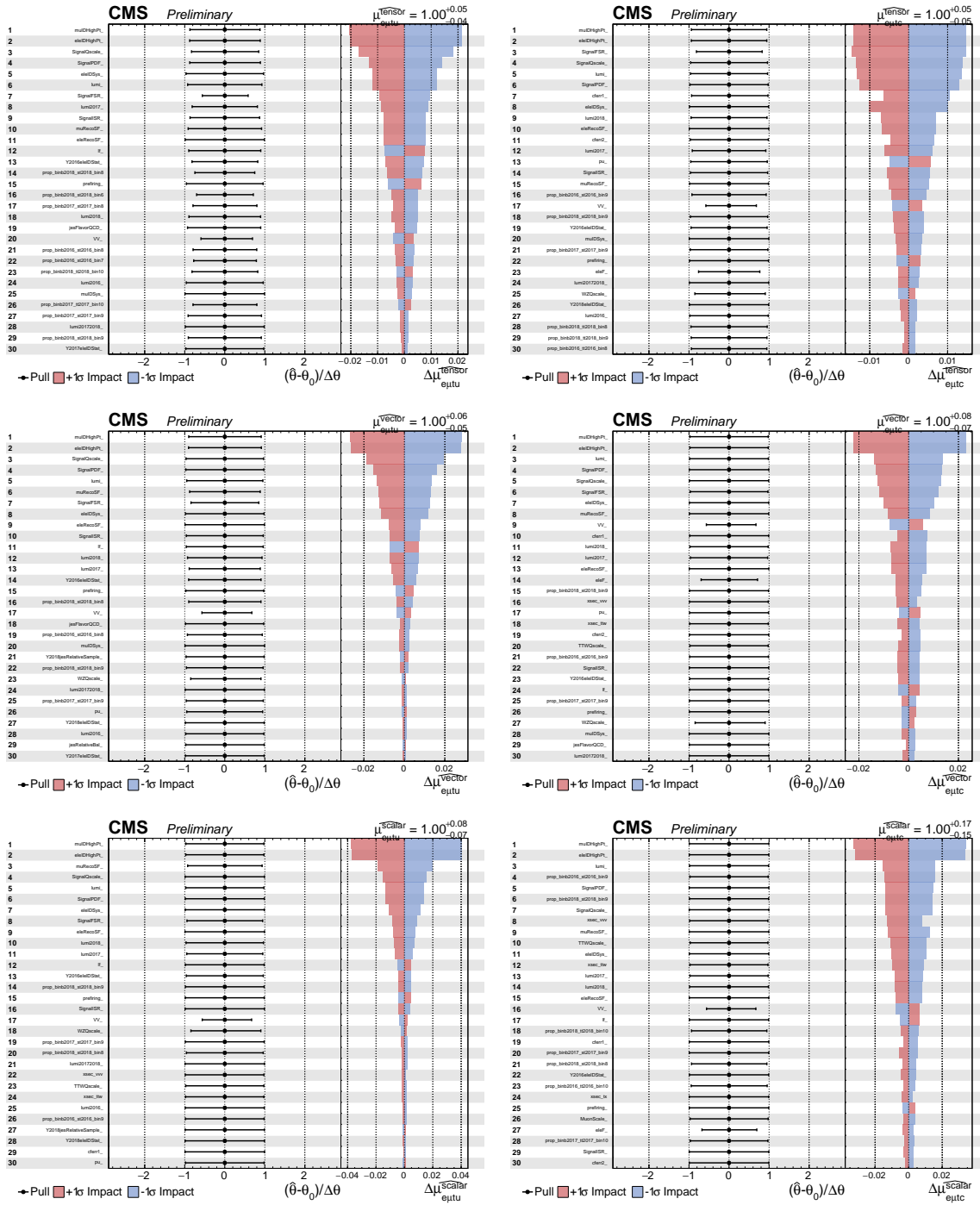


Figure C.3: Expected impact with an expected signal strength at 1. From top to bottom: tensor, vector, scalar. From left to right: $e\mu tu$, $e\mu tc$.