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

ATLAS A Toroidal LHC Apparatus

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

L1 Level-1

MC Monte-Carlo

ME matrix-element

PD Primary Dataset

PDF Parton Distribution Function

SM Standard Model

WC Wilson Coefficient

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CHAPTER	<u> </u>

Introduction

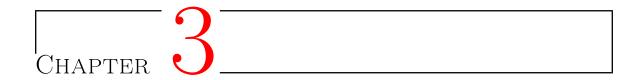
Part I

Theoretical Framework



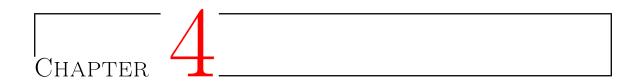
Electroweak Standard Model

- 2.1 Gauge Theory
- 2.2 Higgs Mechanism
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Beyond the Standard Model

- 3.1 Overview of the BSM Theories
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Effective Field Theory

- 4.1 Standard Model EFT
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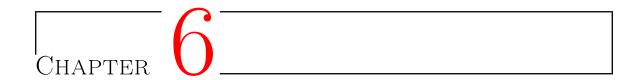


Event Simulation for Colliders

- 5.1 Perturbative Quantum Chromodynamics
- 5.2 Nonperturbative Quantum Chromodynamics

Part II

Experimental Apparatus

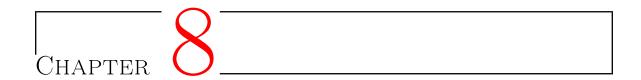


The Large Hadron Collider



The Compact Muon Solenoid Detector

- 7.1 Coordinate System Used in the CMS Detector
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Object Reconstruction in the CMS detector

- 8.1 Electron and Photon
- 8.2 Muon
- 8.3 Tau
- **8.4** Jets
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Run-3 Operation of the CMS detector

- 9.1 The CMS Run Organisation
- 9.2 Central Shifting Crew
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Chapter 10

The Phase-2 Upgrade of the CMS Detector

- 10.1 Overview of the Upgrade
- 10.2 Leve-1 Track Finder
- 10.3 Leve-1 Electron Trigger Algorithm

Part III

Search for Flavor-Violating $e\mu$ tq Interactions

Chapter 11

Previous searches

- 11.1 ATLAS
- 11.2 CMS

[1]

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 in "Primary Datasets (PDs)", which 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. The data taking conditions in 2016-2018 were difference across the years. To account for this, all MC samples are generated separately for each year separately.

12.1 Signal Samples

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

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

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 eutu vertex as an example, the Wilson Coefficients (WCs) are parameterized in the following way:

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

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

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

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

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

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

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

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

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

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

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

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 I and q denote left-handed doublets, whereas u and e denote right-handed singlets. The indices i and jare 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		
	$O_{lq}^{(1)ijmn}$	=	$(\bar{I}_i \gamma^\mu I_j) (\bar{q}_m \gamma_\mu q_n)$
vector	O _{lu}	=	$(\overline{I}_i \gamma^\mu I_j) (\overline{u}_m \gamma_\mu u_n)$
VCCLOI	O_{eq}^{ijmn}	=	$(\overline{\mathbf{e}}_i \gamma^{\mu} \mathbf{e}_j)(\overline{\mathbf{q}}_m \gamma_{\mu} \mathbf{q}_n)$
	O_{eu}^{ijmn}	=	$(\overline{\mathrm{e}}_{i}\gamma^{\mu}\mathrm{e}_{j})(\overline{\mathrm{u}}_{m}\gamma_{\mu}\mathrm{u}_{n})$
scalar	O _{lequ}	=	$(\bar{I}_ie_j)\;\varepsilon\;(\overline{q}_mu_n)$
tensor	$O_{lequ}^{(3)ijmn}$	=	$(\bar{I}_i\sigma^{\mu\nu}e_j)\;\varepsilon\;(\overline{q}_m\sigma_{\mu\nu}u_n)$

Additionally, all vector-like operators are combined,

$$O_{\text{eutu}}^{\text{vector}} = O_{\text{la}} + O_{\text{lu}} + O_{\text{ea}} + O_{\text{eu}}, \tag{12.8}$$

$$\begin{array}{lll} O_{e\mu tu}^{vector} & = & O_{lq} + O_{lu} + O_{eq} + O_{eu}, & (12.8) \\ O_{e\mu tu}^{scalar} & = & O_{lequ}^{(1)} + h.c, & (12.9) \end{array}$$

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

which results in 6 independent WCs: $C_{\mathrm{e}\mu\mathrm{t}u}^{\mathrm{vector}}$, $C_{\mathrm{e}\mu\mathrm{t}u}^{\mathrm{scalar}}$, $C_{\mathrm{e}\mu\mathrm{t}u}^{\mathrm{tensor}}$, $C_{\mathrm{e}\mu\mathrm{t}c}^{\mathrm{vector}}$, $C_{\mathrm{e}\mu\mathrm{t}c}^{\mathrm{scalar}}$, $C_{\mathrm{e}\mu\mathrm{t}c}^{\mathrm{tensor}}$

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 generator MADGRAPH5_AMC@NLO v2.4.2 [6] to generate events at the leading order.

In general, the calculations done by the ME 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-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

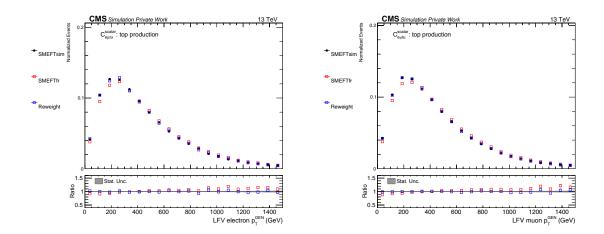


Figure 12.1: Comparison of kinematic distributions at ME-level produced by different models: LFV electron p_T (left), LFV muon p_T (right). The "SmeftFR" samples (shown in red curve) and "SMEFTsim" samples (shown in black curve) are statistically independent of each other. The "Reweight" (shown in blue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

on kinematics, the following ratio is calculated for each event i,

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

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 NNLO precision [8] is used to normalize the top decay signals. The cross sections for all signal MC samples are summarized in Table 12.2.

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 process, the NLO Parton Distribution Function (PDF) set from NNPDF3.0 [9] is used in 2016 to generate background MC samples. The NNLO PDF set from NNPDF3.1 [10] is used for tZq while the LO PDF set from NNPDF3.0 is used for tHq, tHW,

Table 12.2: Theoretical cross sections for top production and decay for each CLFV coupling. Uncertainties related to PDF and QCD scale are given ($\sigma_{-\text{scale}}^{+\text{scale}} \pm \text{PDF}$).

Lorentz Structure	Samples	XS (fb)	
	top production via u quark	$460^{+81}_{-64}\pm 6$	
vector	top production via c quark	$33^{+5}_{-4}\pm 6$	
	top decay via u/c quark	$32^{+0.8}_{-1.1}\pm1.3$	
	top production via u quark	$97^{+18}_{-14}\pm1$	
scalar	top production via c quark	$6.3^{+0.9}_{-0.8}\pm1.4$	
	top decay via u/c quark	$4.0^{+0.1}_{-0.1}\pm0.2$	
	top production via u quark	$2143^{+368}_{-293}\pm31$	
tensor	top production via c quark	$164^{+22}_{-18}\pm27$	
	top decay via u/c quark	$187^{+5}_{-6}\pm 8$	

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 generator at CMS is MADGRAPH5_AMC@NLO v2.4.2 (v2.2.2 for 2016), which is used to generate all but ZZ, ttH, and tt samples. These three samples are generated with POWHEG v2 [11] instead. The PYTHIA v8.2 [12] is used to model parton shower and hadronization. The CUETP8M1 [13] is used in 2016 for underlying event tuning while the CP5 [14] is used in 2017 and 2018. The configurations of the MC samples are summarized in Table 12.3. The nonprompt backgrounds in this analysis will be modeled with data-driven technique. The MC samples listed in "nonprompt" prompt category of Table 12.3 are only used as reference for initial studies.

12.3 Triggers

To achieve a trigger optimal acceptance, a combination of single-lepton, di-lepton and tri-lepton triggers are used to select events. These triggers are summarized in Appendix A. The following trigger logic was implemented to remove the overlap between different datasets:

- Events in MC datasets are required to fire at least one of the triggers listed in Table A.1-A.3.
- Events in Single Muon datasets are required to fire at least one of the triggers listed under "SingleMuon".
- Events in Double Muon 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".

Category	Process	Event Generator	Perturbative QCD	Tune	XS precision
	WZ	Madgraph	NLO	CUETP8M1(CP5)	NLO [15]
	ZZ	Powheg	NLO	CUETP8M1(CP5)	NLO [15]
_	VVV	Madgraph	NLO	CUETP8M1(CP5)	NLO
Prompt backgrounds	$t\bar{t}W$, $t\bar{t}Z$	Madgraph	NLO	CUETP8M1(CP5)	NLO [16, 17]
J	tŧH	Powheg	NLO	CUETP8M1(CP5)	NLO [17]
	tZq	Madgraph	NLO	CP5	NLO
	tHq, tHW, tWZ	Madgraph	LO	CUETP8M1(CP5)	LO
	tŧ	Powheg	NLO	CUETP8M1(CP5)	NNLO [8]
Non-prompt backgrounds	DYM50	Madgraph	NLO	CUETP8M1(CP5)	NNLO [18]
9	DYM10to50	Madgraph	LO	CUETP8M1(CP5)	NLO [18]

Table 12.3: Summary of the configurations of the MC samples.

- 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

13.1 Leptons

Electron

We apply the following selections to electron objects:

- $p_T > 20$ GeV,
- $|\eta|$ < 2.4,
- ullet the gap (1.4442 < $|\eta_{SC}|$ <1.566) is removed,
- ullet $|d_{xy}|$ < 0.05 cm, $|d_z|$ < 0.1 cm,
- missing inner hits < 2,
- ullet significance of the 3D impact parameter SIP $_{3D}$ <8,
- pass the TOP MVA ID Tight WP citemvaTOP (> 0.9 across all years),
- minilsolation < 0.12.

Electron energy scale corrections citeEcalScale are applied.

Muon

Similarly, we apply the following selections to muon objects:

- $p_T > 20$ GeV,
- $|\eta|$ < 2.4,
- $ullet \; |d_{{\scriptscriptstyle X}{\scriptscriptstyle Y}}| \; < 0.05 \; {
 m cm}, \; |d_{{\scriptscriptstyle Z}}| \; < 0.1 \; {
 m cm},$
- \bullet significance of the 3D impact parameter SIP_{3D} <8,
- pass the cut-based muon ID citemuID (medium working point),
- pass the TOP MVA ID Tight WP citemvaTOP (> 0.9 across all years),
- minilsolation < 0.12.

Muon energy (Rochester) corrections are applied to low p_T muons ($p_T < 200 \text{GeV}$).

The TOP MVA ID is used to suppress the contamination of non-prompt leptons. Both electron and muon selection in the analysis is synchronized with.

The isolation variable "minilsolation" used to select both electrons and muons is also used as input to the TOP MVA training. It is similar to a normal isolation variable but with a pT dependent cone size:

$$R = \max(0.05, \min(0.2, \frac{10 \, GeV}{p_T})). \tag{13.1}$$

This analysis requires three leptons selected by the criteria above. The p_T threshold on the leading lepton is 38 GeV.

13.2 **Jets**

Jets are reconstructed from particle flow, PF, candidates using the Anti-kT clustering algorithm with a cone parameter R=0.4 (AK4). The Charged Hadron Subtraction algorithm, CHS, is applied to the jets, AK4PFCHS jets, in order to groom them, removing extra radiation from pileup. After this preliminary reconstruction the groomed jets must satisfy the below conditions:

- $p_T > 30$ GeV,
- $|\eta|$ < 2.4,
- TightLepVeto citejetID,
- Pile up jet ID loose citepujetID, if p_T < 50 GeV.

Jets are also required to be isolated from selected leptons. A cone of Δ R < 0.4 around each jet candidate is defined and jets are removed if any selected leptons are found within such a cone.

Jets from MC are corrected in order to match the data which is skewed by pileup (L1FastJet) and L2L3 MC-truth corrections. In order to correct the MC jets, we applied the recommended Jet Energy Corrections. For reference the versions of the JECs we used for

13.3 B-tagging

Chapter 14

Event Selection

We require events to contain exactly three selected leptons. Events with different lepton compositions are further categorized into different channels: eee, $\mu\mu\mu$, $e\mu l$. In all three channels, the sum of the charges of three leptons are required to be 1 or -1. The leading leptons in all selected events are required to be matched with trigger objects via $\Delta R < 0.2$. Within each channel, different regions are defined to further understand signal and background.

 $e\mu\ell$ is the channel where the presence of the signal is expected. We require at least one pair of leptons of Opposite-Sign-Opposite-Flavor (OSOF). Such a requirement is achieved if:

- Both positively charged and negatively charged leptons are present.
- Both electrons and muons are present.

Events with the presence of the same flavor leptons are categorized into eee or $\mu\mu\mu$ channels to study the background composition.

14.1 Signal Region

In $e\mu l$ channel, we set up our signal region (SR) targeting LFV signal processes. Apart from the OSOF pair, we require the events to have big MET (MET > 20 GeV). Events with an Opposite-Sign-Same-Flavor (OSSF) lepton pair, the invariant mass of which is between 50 GeV and 106 GeV, are removed to suppress contribution from Z production. To further suppress background contribution, we require at least one jet and at most one b-tag jet.

Distributions of leading lepton p_T and invariant mass of Opposite-Sign $e\mu$ pair are shown in Figure 14.1. Please note: Although we intend to deploy a data-driven technique to estimate non-prompt background in the actual analysis, all backgrounds in Figure 14.1 are estimated using MC simulation only. This serves as a preliminary check to understand the components of different backgrounds in SR.

Using the invariant mass of the Opposite-Sign $e\mu$ pair, SR is further divided into two subsets to create ST and TT enriched regions:

• SR1, $M_{e\mu}$ <150GeV: TT enriched.

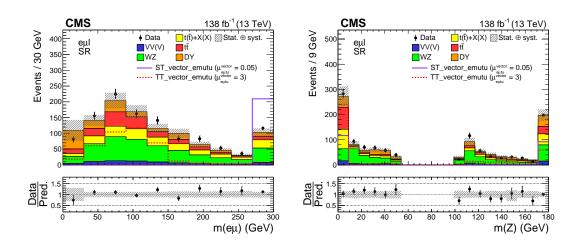


Figure 14.1: Distributions of different kinematic variables estimated in SR (full run II). Backgrounds are estimated using MC only. From left to right: invariant mass of the $e\mu$ pair, leading lepton p_T .

• SR2, $M_{e\mu}$ >150GeV: ST enriched.

14.2 Validation Region

Events are required to pass trigger requirements listed in Table A.1-A.3. The leading lepton is required to have a $p_T > 38$ GeV. The sum of the charges of the three selected leptons is required to be either 1 or -1. Selections applied to different regions are summarized in Table 14.1 and are illustrated in Figure 14.2.

Acronyms

- $-C_i$: electric charge of the i-th lepton
- OnZ: requiring the presence of at least one Opposite-Sign-Same-Flavor lepton pair with an invariant mass between 50 GeV and 106 GeV.
- OffZ: veto events that pass OnZ
- MET: missing transverse energy
- njet: number of jets that pass jet selection
- nbjet: number of b-tag jets that pass jet selection
- SR: signal region
- VR: validation region for data-driven method
- WZ CR: WZ control region

Channel	Region	$ \Sigma_i C_i = 1$	OnZ	OffZ	MET > 20 GeV	njet>=1	nbjet<=1
eee	VR	\checkmark	-	-	-	-	-
	WZ VR	\checkmark	\checkmark	-	\checkmark	\checkmark	\checkmark
еμℓ	SR	✓	-	✓	√	√	✓
	Nonprompt VR	\checkmark	\checkmark	-	-	-	-
	WZ VR	\checkmark	\checkmark	-	\checkmark	\checkmark	✓
μμμ	Nonprompt VR	\checkmark	-	-	-	-	-
	WZ VR	\checkmark	-	\checkmark	\checkmark	\checkmark	\checkmark

Table 14.1: Summary of the cuts applied to different regions

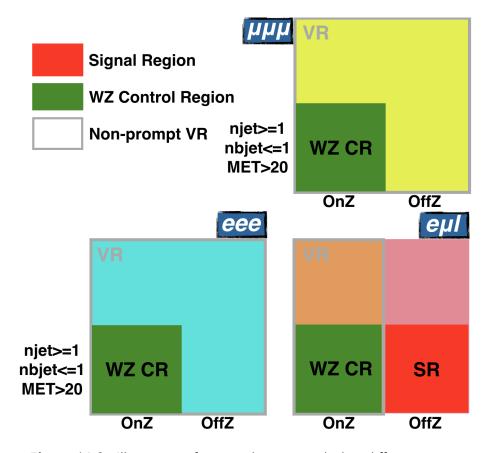


Figure 14.2: Illustration of event selections applied to different regions

There are two types of signal-depleted regions defined within eee/ $\mu\mu\mu$ channel: validation region (VR) and WZ control region. The distinction between these two regions comes down to

background compositions. It is expected that the VR has a significant fraction of non-prompt backgrounds while WZ production is responsible for most of the backgrounds in the WZ control region. Distributions of leading lepton p_T and leading lepton η in WZ control region can be found in Figure 14.3-14.5. Discussion about validation region can be found in chapter 15.

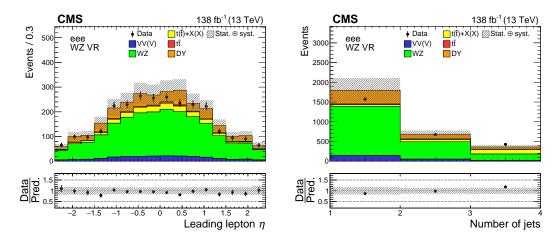


Figure 14.3: Distributions of different kinematic variables estimated in WZ control region (full run II). Backgrounds are estimated using MC only. From left to right: leading lepton η , jet multiplicity.

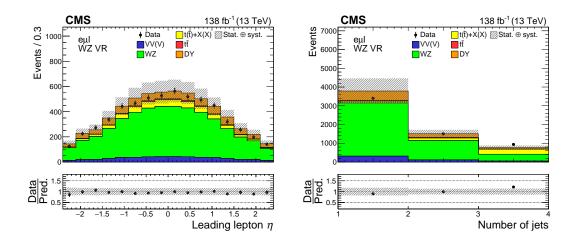


Figure 14.4: Distributions of different kinematic variables estimated in WZ control region (full run II). Backgrounds are estimated using MC only. From left to right: leading lepton η , jet multiplicity.

14.3 Kinematic Reconstruction

A kinematic reconstruction is performed in this channel. The OSOF lepton pair is assumed to be the product of cLFV interaction, while the third lepton (bachelor lepton) is assumed to be the product of the SM top-quark decay. Jet with the highest b-tagging score is assumed to originate from bottom-quark decay, therefore, it is combined with MET and the bachelor lepton to build

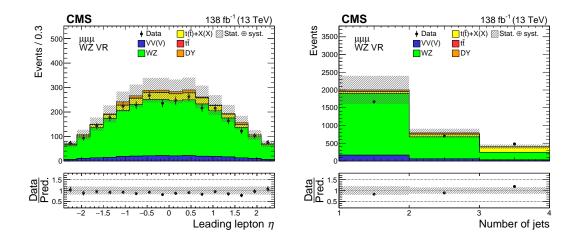


Figure 14.5: Distributions of different kinematic variables estimated in WZ control region (full run II). Backgrounds are estimated using MC only. From left to right: leading lepton η , jet multiplicity.

the SM top-quark.

- We take the x and y component of MET 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 (bachelor lepton+neutrino) must be equal to W boson mass.
- If there is no real solution, we take the real part of the complex solution.
- If there is more than one real solution, we chose the solution that is the closest to the p_z of the bachelor lepton.

In events where there is more than one candidate of bachelor lepton, the lepton that gives a top mass that is the closest to the the SM top-quark mass is chosen as the bachelor lepton.

Once the bachelor lepton has been determined, the OSOF pair is combined with jets One-By-One to reconstruct LFV top 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 candidates, the candidate that gives a top mass that is the closest to the the SM top-quark mass is chosen.

Kinematic reconstruction of heavy objects like top-quark is carried out using selected leptons, jets, and MET.

• Z boson candidate

 Z boson candidate is reconstructed by combining two Opposite-Sign-Same-Flavor leptons Whenever possible, Z candidate with an invariant mass that is closest to Z boson mass is chosen

Standard model top-quark candidate

- Standard model top-quark candidate is reconstructed by combining a jet, a lepton, and MET
- Jet with the highest b-tagging score is always chosen
- In $e\mu I$ channel, flavor-violating leptons are excluded from reconstruction
 - * Whenever more than one lepton is available due to possible combinatorics of flavor-violating $e\mu$ pair, lepton that forms a top-quark candidate with an invariant mass that is closest to the top-quark mass is chosen
- In $eee/\mu\mu\mu$ channel, leptons that form Z candidate are excluded
- Lepton is combined with MET to reconstruct W candidate
 - * Transverse component of the neutrino momentum is taken from MET
 - * Neutrino p_z is obtained by W mass constrain set on the lepton-neutrino system

• Flavor-violating top-quark candidate

- Flavor-violating top-quark candidate is reconstructed by combining a jet and an Opposite-Sign $e\mu$ lepton pair
- Jet with the highest b-tagging score is excluded from reconstruction unless it fails b-tagging
- Whenever possible, the jet that forms a flavor-violating top-quark candidate with an invariant mass that is the closest to the top-quark mass is chosen
- Lepton that forms the standard model top-quark candidate is excluded

• The first b-jet+lepton system

- Jet with the highest b-tagging score is always chosen
- All three leptons can form this system with the chosen jet
 - * Lepton that forms a jet-lepton system with the lowest invariant mass is chosen

• The second b-jet+lepton system

- Jet with the highest b-tagging score is excluded unless it passes b-tagging
- Whenever possible (after the previous step), jet with the highest b-tagging score is chosen
- Lepton that forms the first jet-lepton system is excluded

- Leptons that have the same sign as the lepton that forms the first jet-lepton system is excluded
- Whenever possible, lepton that forms a jet-lepton system with the lowest invariant mass is chosen

CHAPTER 15

Nonprompt Background Estimate

In this analysis, the term *real* leptons refers to prompt leptons that originate from the primary interaction, like the decays of vector bosons that are produced via hard collisions. *Fake* leptons refer to non-prompt leptons that are mostly produced in the decays of b-quarks, c-quarks, pions, or from photon conversion. A relatively low ratio of *fake* leptons in a lepton selection is desired and can be achieved by applying stringent isolation requirements and/or an MVA-based lepton identification specifically trained against *fake* leptons.

Non-prompt backgrounds are defined to be backgrounds with at least one *fake* lepton passing the analysis selection, in this case generally dominated by Drell-Yan and $t\bar{t}$ production. An accurate estimation of non-prompt backgrounds is difficult to achieve through MC modeling. Therefore, a data-driven technique called the 3D matrix method is used to estimate the non-prompt backgrounds.

A brief description of the matrix method is given in this section followed by details on it's implementation. In autorefyou can find more information on the generalization of the matrix method to 2D and 3D. The appendices also contain a validation of the method and its performance is compared to the results of the fake factor method, another commonly used data-driven technique.

15.1 The Matrix Method

The matrix method citematrix2012 is a data driven technique used to estimate the fraction of fake leptons that pass a given lepton selection, referred to as "tight". The tight selection usually incorporates tight lepton identification and isolation requirements and should be similar if not exactly the same as lepton selection in the actual analysis. The loose selection is obtained by loosening the tight selection. The loose selection is used as a baseline such that any leptons that pass it will fall into one of the two exclusive categories: tight or not tight. The matrix method treats real and fake leptons separately. As a result, real and fake efficiencies are introduced (see Figure 15.1).

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

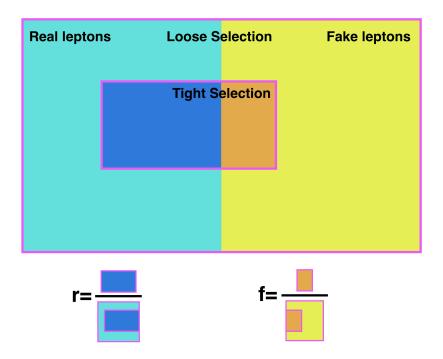


Figure 15.1: Illustration of the *real* efficiency r and the *fake* efficiency f.

through measurement:

$$r = \frac{n_R^T}{n_R^T + n_R^{\overline{T}}},\tag{15.1}$$

in which n_R^T/n_R^T denotes the number of events with a *real* lepton that is *tight/not tight*.

Similarly, fake efficiency f can be expressed as

$$f = \frac{n_F^T}{n_F^T + n_F^{\overline{T}}},\tag{15.2}$$

in which n_F^T/n_F^T denotes the number of events with a fake lepton that is $tight/not\ tight$.

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

In our 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:

$$N^{T} = N_{R}^{T} + N_{F}^{T}$$

$$N^{\overline{T}} = N_{R}^{\overline{T}} + N_{F}^{\overline{T}},$$
(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. $N_R^{\overline{T}}/N_F^{\overline{T}}$ can be expressed in terms of r/f and N_R^T/N_F^T according to Equation 15.1/15.2 and the assumption that r/f is transferable to other regions:

$$N^{T} = r \frac{N_{R}^{T}}{r} + f \frac{N_{F}^{T}}{f}$$

$$N^{\overline{T}} = (1 - r) \frac{N_{R}^{T}}{r} + (1 - f) \frac{N_{F}^{T}}{f}.$$
(15.4)

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

$$\begin{pmatrix} N^T \\ N^{\overline{T}} \end{pmatrix} = \begin{pmatrix} r & f \\ 1 - r & 1 - f \end{pmatrix} \begin{pmatrix} N_R^T/r \\ N_F^T/f \end{pmatrix}.$$
(15.5)

Since those two numbers that appear on the lefthand side of the equation can be obtained through measurement, a matrix inversion can give us an estimation of *fake* leptons that pass tight selection, denoted by N_F^T .

15.2 Extraction of Prompt and Nonprompt Efficiencies

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 (i.e. 3D matrix method. See autorefsec:appendixnonprompt for more details), where all three leptons are considered to be possibly *fake*.

The measurement of both r and f are performed in dilepton MRs using the Tag-and-Probe approach. Datasets used to measure r and f are listed in Table. More information about MC samples listed in Table ??-. More can be found. Events are required to fire either single-electron or single-muon trigger listed in Table 15.1.

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,

Channel	path	dataset	2016	2017	2018
Electron	HLT_Ele27_WPTight_Gsf	data & MC	✓	-	-
	HLT_Ele35_WPTight_Gsf	data & MC	-	\checkmark	✓
Muon	HLT_IsoMu27	data & MC	\checkmark	\checkmark	\checkmark

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

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 jet, at least 1 jet}.

The addition of jet multiplicity as a dependency is motivated by the unideal assumption that both r and f are independent of physics processes. We know that fake leptons of different origins are often reconstructed at a different rate. For instance, fake electrons coming from the decays of b-quarks have very different kinematics than fake electrons that originate from photon conversion. Therefore, the probabilities of these two types of fake electrons to pass any kinematic-dependent selections are going to differ. In other words, the proportion of $t\bar{t}$ events that constitutes the non-prompt backgrounds has a direct impact on r and f. To achieve a better representation of the target region (in terms of background composition), r and f are linked to the number of jets in events.

The *fake* efficiency is measured in a same-sign dilepton region, in which the leading p_T lepton, used as the tag (See Table 15.2 for definition), is required to be matched with trigger objects via $\Delta R < 0.2$. The sub-leading lepton is required to pass the *loose* selection and is taken as the *probe*. Two orthogonal regions are defined within the dilepton MR, which account for the two bins of jet multiplicity:

- Events in dilepton MR with at 0 jets.
- Evens in dilepton MR with at least 1 jet, and at most 1 b-tag jet.

The selection of the second MR is motivated by the selection of the SR, in which at least one jet and at most one b-tag jet are required.

The contribution from prompt backgrounds, estimated from MC simulation, are subtracted from data (see Figure 15.2). Therefore, the fake efficiency f is calculated as:

$$f = \frac{n_{data}^{T} - n_{Prompt\ MC}^{T}}{n_{data}^{T+\overline{T}} - n_{Prompt\ MC}^{T+\overline{T}}}.$$
(15.6)

Lepton	Selection	Loose	Tag	<i>Tight^b</i> (Passing-Probe)
Electron	PΤ	> 20 GeV	> 38 GeV ^a	> 20 GeV
	Įmini iso	< 0.4	< 0.1	< 0.12
	TOP MVA ID	>-0.9	>0.95	>0.9
	Match with trigger objects	-	\checkmark	-
Muon	PΤ	> 20 GeV	> 30 GeV	> 20 GeV
	Imini iso	< 0.4	< 0.1	< 0.12
	cut-Based ID	-	Medium WP	Medium WP
	TOP MVA ID	>0.5	>0.9	>0.9
	Match with trigger objects	-	\checkmark	-

Table 15.2: Summary of the lepton selections needed for the measurement of r and f. Please note: a) the minimum p_T cut for tag electron in 2016 dataset is reduced to 30 GeV to adjust for the trigger threshold; b) the tight selection here is the same as the lepton selection listed in autorefsec:reconstruction.1.

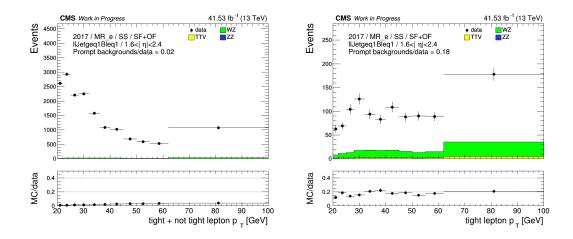
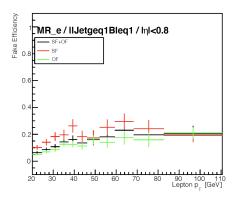


Figure 15.2: Distribution of lepton p_T in an electron *fake* efficiency MR (2017). In this particular example: both e+e (Same-Sign-Same-Flavor) and $e+\mu$ (Same-Sign-Opposite-Flavor) pair are considered. At least one jet and at most one b-tag jet are required (the second jet multiplicity bin). *Probe* electron is required to have $1.6 < |\eta| < 2.4$ (the third η bin). Contamination from prompt backgrounds are estimated with MC simulation. From left to right: $tight+not\ tight$ electron, only tight electron.

The dependency of f on background composition is also reflected in the difference between f measured with different flavor compositions. For electrons, the *fake* efficiency measured with $e + \mu$ selection is visibly different than the *fake* efficiency measured with e + e selection, which



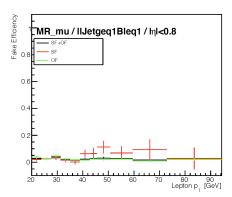


Figure 15.3: Fake efficiency measured in 2017 data. These plots correspond to the first $|\eta|$ bin $(|\eta| < 0.8)$ and the second jet multiplicity bin. Error bars displayed in these plots include statistical uncertainty only. From left to right: electron f, muon f.

could suggest:

- Fake electrons that originate from the decays of b-quarks have a lower probability of being selected by *tight* selection than *fake* electrons coming from Z+jets events.
- The Same-Sign e + e selection is contaminated with *real* electrons and, to a less extent, the Same-Sign $e + \mu$ selection.

Events that have two same-sign electrons with an invariant mass between 76 GeV and 106 GeV are removed from MR to suppress the charge-flip effect of electrons. No such cut has been introduced to muon MR due to the negligible rate of charge-flip in muons.

Lepton selections listed in Table 15.2 are referred to as *baseline*, which has been optimized to provide a robust non-prompt estimate. Variation of the *basline* selection has also been studied.

There is a trade-off between muon r and f when changing the isolation requirement (See Figure 15.4-15.5). While it is ideal to reduce statistical uncertainties by loosening the *loose* definition, a low *real* efficiency should also be avoided. This change proves to have a minor effect on the overall estimate.

The *real* efficiency r is estimated with simulated $t\bar{t}$ events in an Opposite-Sign dilepton region. The same lepton selection (See Table 15.2) is used to perform Tag-and-Probe. The leading lepton in p_T is used as a tag while the sub-leading lepton (with opposite sign) is taken as a probe. The variation of r between e+e and $e+\mu$ flavor combination is negligible (See Figure 15.6). Therefore, only $e+\mu$ events are used to measure real efficiency in order to minimize the contamination of fake leptons.

The selection criteria for r/f measurement region is summarized in Table 15.3. The latest version

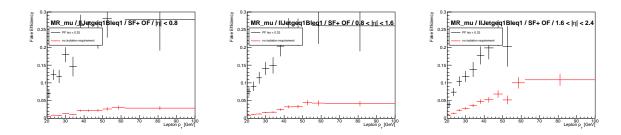


Figure 15.4: Comparison of the muon *fake* efficiencies measured with isolation requirement in *loose* definition (2017 datasets). From left to right: $|\eta| < 0.8$, $0.8 < |\eta| < 1.6$, $1.6 < |\eta| < 2.4$.

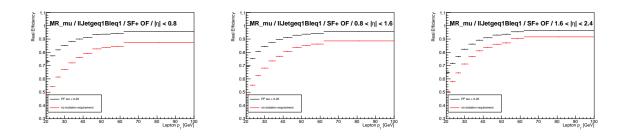


Figure 15.5: Comparison of the muon *real* efficiencies measured with different *loose* definition (2017 datasets). From left to right: $|\eta| < 0.8$, $0.8 < |\eta| < 1.6$, $1.6 < |\eta| < 2.4$.

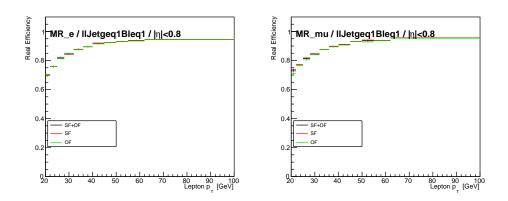


Figure 15.6: Real efficiency measured in simulated $t\bar{t}$ events (2017). These plots correspond to the first $|\eta|$ bin ($|\eta|$ <0.8) and the second jet multiplicity bin. Error bars displayed in these plots include statistical uncertainty only. From left to right: electron r, muon r.

of the real and fake efficiency is summarized in autorefsec:RandF.

Observable	jet bin	# of selected leptons	lepton flavor composite	$ \Sigma_i C_i $	OffZ	njet>=1	nbjet<=1
f	0 jet	2	any	1	✓	-	-
1	1 or more jet	2	any	1	\checkmark	\checkmark	\checkmark
	0 jet	2	$e\mu$ only	0	✓	-	-
1	1 or more jet	2	$e\mu$ only	0	\checkmark	✓	\checkmark

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

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 with MC simulation while non-prompt backgrounds are estimated with the matrix method. A summary of the selections applied to these VRs is given in autorefsec:selection.4.

Distribution of lepton lepton η and jet multiplicity are shown in Figure 15.7-15.9. Distributions of other variables are included in autorefsec:VRmatrix

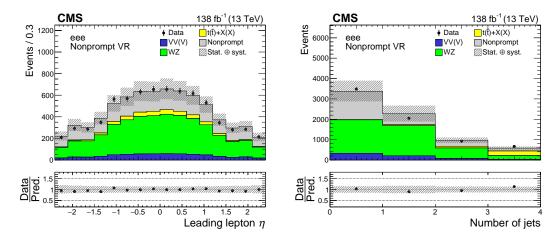


Figure 15.7: Distributions of different kinematic variables estimated in VR with three electrons. From left to right: leading lepton η , jet multiplicity.

15.4 Nonprompt estimate in SR

The matrix method is used to estimate non-prompt backgrounds in the signal region due to its superior performance over other alternative methods (e.g. fake factor method discussed in autorefsec:appendixnonprompt). Distributions of leading lepton η , jet multiplicity, n-jet multiplicity, missing transverse energy, LFV-Top mass and Z mass are shown in Figure 15.10-15.11. Distributions of other variables are included in autorefsec:SRDataDriven.

The number of expected events from various kinds of backgrounds are shown in Table 15.4.

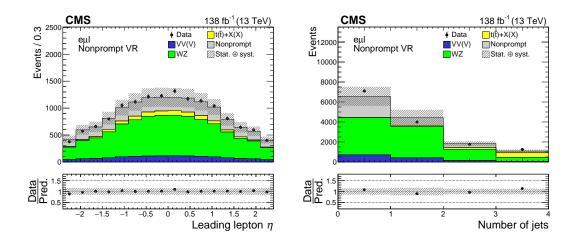


Figure 15.8: Distributions of different kinematic variables estimated in VR with electron, muon and a third light lepton. From left to right: leading lepton η , jet multiplicity.

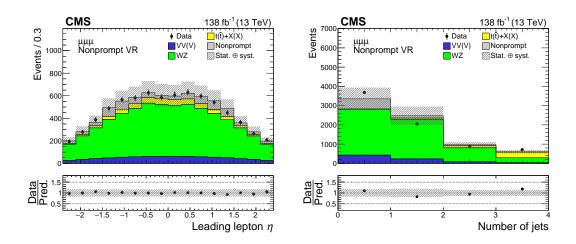


Figure 15.9: Distributions of different kinematic variables estimated in VR with three muons. From left to right: leading lepton η , jet multiplicity.

Representative bin yields and their statistical uncertainties are summarized in citeStatError.

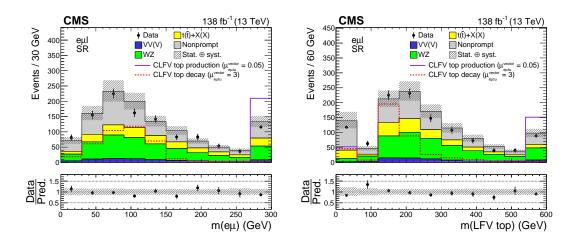


Figure 15.10: Distributions of different kinematic variables estimated in SR (full run II). From left to right: b-jet multiplicity, missing transverse energy.

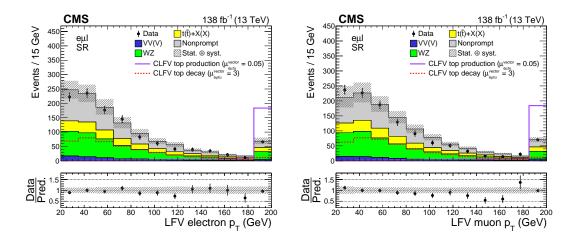


Figure 15.11: Distributions of different kinematic variables estimated in SR (full run II). From left to right: b-jet multiplicity, missing transverse energy.

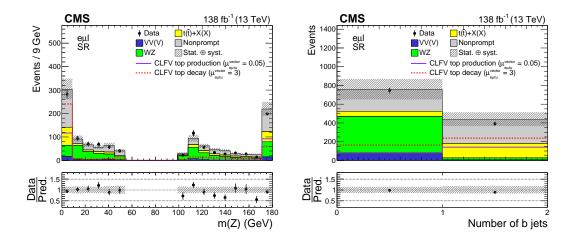


Figure 15.12: Distributions of different kinematic variables estimated in SR (full run II). From left to right: mass of the flavor-violating-top-quark candidate, mass of the Z boson candidate

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 $C_{\rm e\mu tu}^{\rm vector}/\Lambda^2=1 {\rm TeV}^{-2}$, is also listed for reference. The signal yields include contributions from both top production and decay modes.

Process	m(eμ)<150 GeV	m(eμ)>150 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
tīZ	39.1 ± 7.9	25.8 ± 5.4
tŧH	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

Chapter 16

Signal Extraction with Boosted Decision Trees

To further separate the LFV signal from the standard model backgrounds in our signal region, a Boosted Decision Tree (BDT) was employed using the xgBoost package citeChen:2016:XST:2939672.2939785.

16.1 BDT Setup

Since the kinematic distributions of final state particles in our two signal channels (single top production and top decay) differ significantly (see Figure 16.3-16.7), the SR is divided into two parts (SR1: $M_{e\mu} \leq 150 \, \text{GeV}$; SR2: $M_{e\mu} > 150 \, \text{GeV}$) in order to achieve the optimal sensitivity. Within each sub-region of the SR, a binary BDT is trained.

The signal datasets used for BDT training are MC samples that correspond to single top LFV production (ST) and top LFV decay (TT). The Lorentz structure and the flavor of the up-type quark involved in LFV interaction were shown to have a minor impact on the kinematics of final state particles (See Figure 16.1). Therefore, they were not further distinguished.

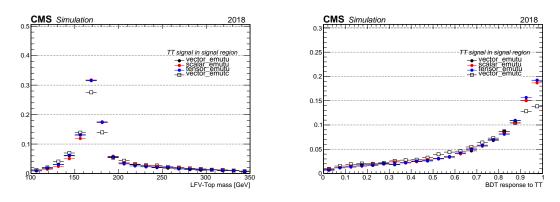


Figure 16.1: Normalized distribution in SR1. From left to right: LFV top mass, BDT shape

The background MC samples include all the samples listed in Table ??-??. It was determined that splitting the background into separate classes (prompt vs non-prompt backgrounds) did not increase the signal discrimination power and caused overtraining which is why in the backgrounds are combined.

The MC dataset is divided into 5 subsets evenly. A 5-fold cross-validation scheme is deployed:

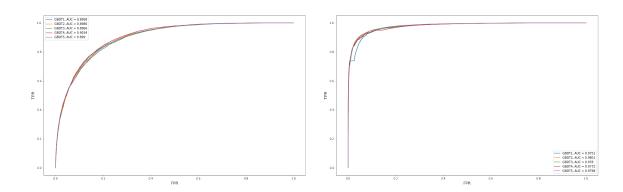


Figure 16.2: ROC curve with 5-fold cross validation. From left to right: BDT targeting TT (SR1), BDT targeting ST (SR2).

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.

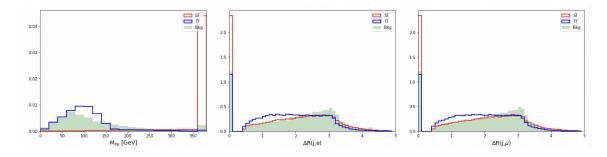


Figure 16.3: Normalized distribution of various features in SR. From to left to right: MVA $_$ M $_{e\mu}$, MVA $_$ JeDr, MVA $_$ JeDr.

16.2 Input Variables

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 IIDr: 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.

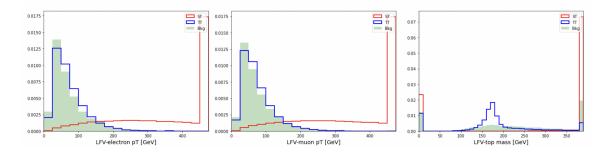


Figure 16.4: Normalized distribution of additional features in SR. From to left to right: MVA LFVePt, MVA LFVmuPt, MVA LFVTopmass.

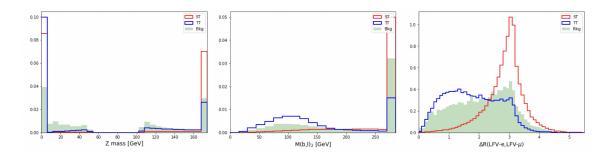


Figure 16.5: Normalized distribution of additional features in SR. From to left to right: MVA Zmass, MVA Mbl2, MVA IIDr.

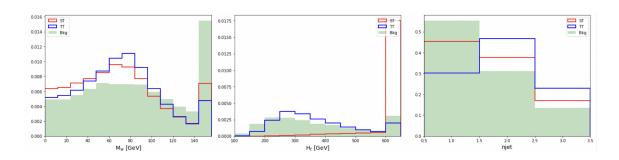


Figure 16.6: Normalized distribution of additional features in SR. From to left to right: MVA_tM, MVA Ht, MVA njet.

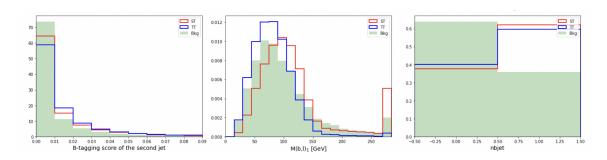


Figure 16.7: Normalized distribution of additional features in SR. From to left to right: MVA Jet2Btag, MVA Mbl1, MVA nbjet.

16.2.1 Input ranking and correlation

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

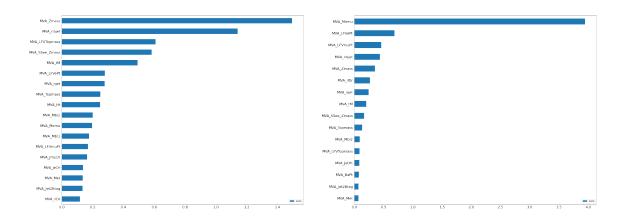


Figure 16.8: List of features ranked by their relative importance. From left to right: BDT targeting TT (SR1), BDT targeting ST (SR2)

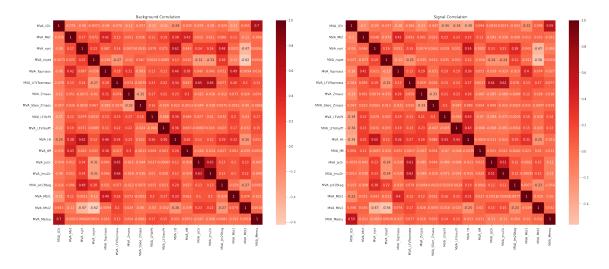


Figure 16.9: Correlation matrices (SR1), from left to right: background correlation, signal correlation.

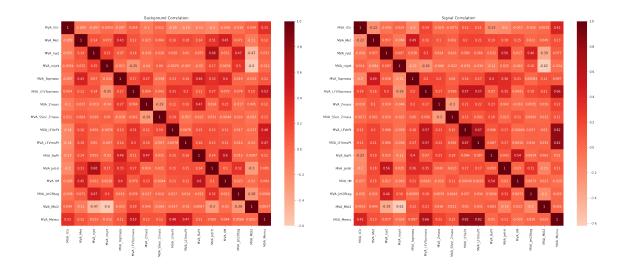


Figure 16.10: Correlation matrices (SR2), from left to right: background correlation, signal correlation.

16.2.2 Input validation

The background modeling of the input features are checked in SR. The input features are shown in Figure 16.11-16.17. Note: all input samples come from MC simulation.

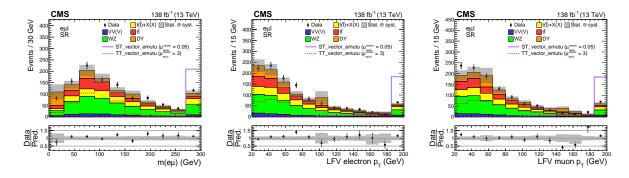


Figure 16.11: BDT input validation, from left to right: MVA_M $_{e\mu}$, MVA_LFVePt, MVA LFVmuPt

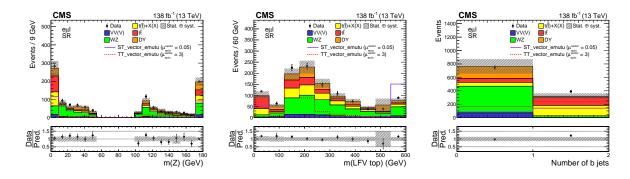


Figure 16.12: BDT input validation, from left to right: MVA_Zmass, MVA_LFVTopmass, MVA nbjet

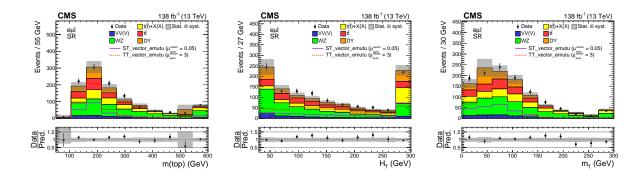


Figure 16.13: BDT input validation, from left to right: MVA_Topmass, MVA_JetHt, MVA_tM

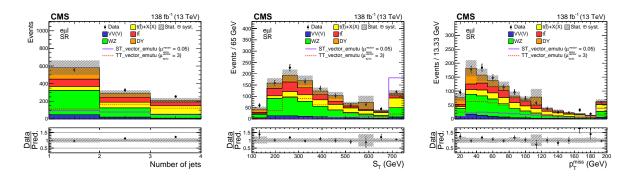


Figure 16.14: BDT input validation, from left to right: MVA njet, MVA Ht, MVA Met

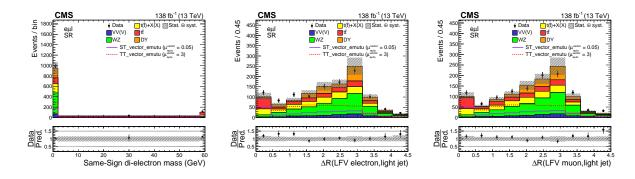


Figure 16.15: BDT input validation, from left to right: MVA_SSZmass, MVA_JeDr, MVA_JmuDr

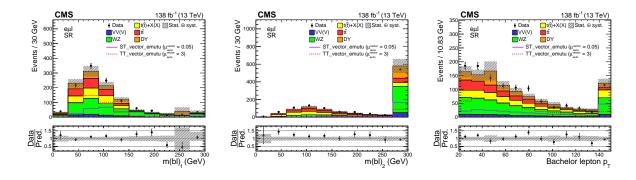


Figure 16.16: BDT input validation, from left to right: MVA Mbl1, MVA Mbl2, MVA BaPt

16.3 BDT Output

The output of the BDTs are shown in Figure 16.18. Note: all but the nonprompt backgrounds are estimated with MC simulation. The nonprompt background backgrounds are estimated with the matrix method.

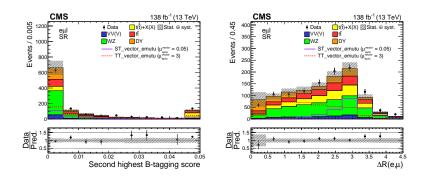


Figure 16.17: BDT input validation, from left to right: MVA Jet2Btag, MVA IIDr

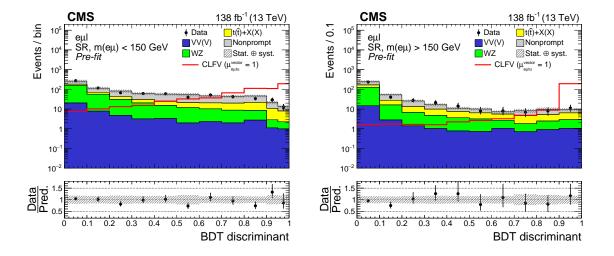
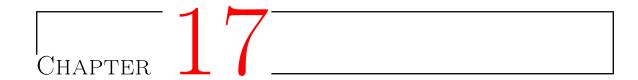


Figure 16.18: Distribution of BDT output, from left to right TT enriched SR (SR1), ST enriched SR (SR2)



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.

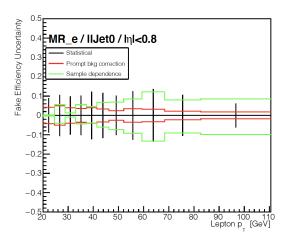
17.2 Nonprompt Uncertainties

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 ??, 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 %.

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



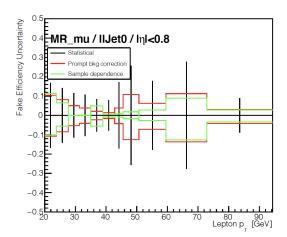


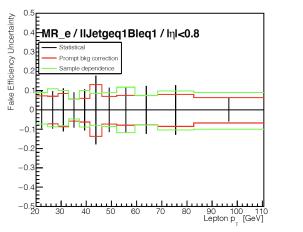
Figure 17.1: Comparison of different components of the uncertainties associated to the *fake* efficiency measured in 2017 dataset (njet=0 bin, $|\eta| < 0.8$ bin). From left to right: electron f uncertainty, muon f uncertainty. Note: black uncertainty bars represent statistical uncertainties.

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):

$$f_{e} = \frac{(1+\beta)n_{F,e+e}^{T} + (1-\beta)n_{F,e+\mu}^{T}}{(1+\beta)n_{F,e+e}^{T+\overline{T}} + (1-\beta)n_{F,e+\mu}^{T+\overline{T}}}.$$
(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.



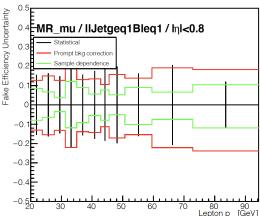
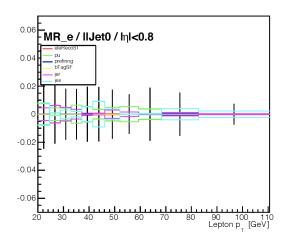


Figure 17.2: Comparison of different components of the uncertainties associated to the *fake* efficiency measured in 2017 dataset (njet>0 bin, $|\eta|$ <0.8 bin). From left to right: electron f uncertainty, muon f uncertainty. Note: black uncertainty bars represent statistical uncertainties.

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 rewighting
 - ECAL prefiring

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.



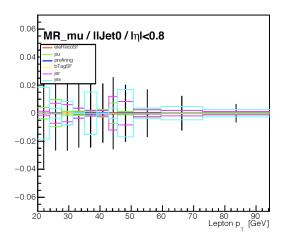


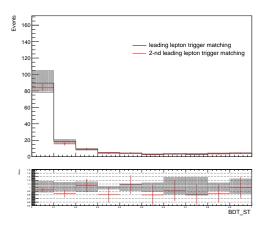
Figure 17.3: Comparison of different components of the uncertainties associated to the *real* efficiency measured in 2017 dataset (njet=0 bin, $|\eta| < 0.8$ bin). From left to right: electron r uncertainty, muon r uncertainty. Note: black uncertainty bars represent statistical uncertainties.

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 ?? (i.e. $N^{\overline{T}TT}$, $N^{\overline{T}TT}$, $N^{\overline{T}TT}$, $N^{\overline{T}TT}$) 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 trilepton 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



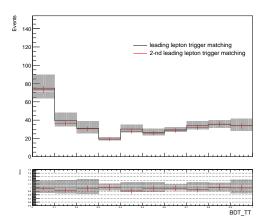


Figure 17.4: The impact of matching leptons to trigger objects on non-prompt estimate. From left to right: non-prompt estimate in ST enriched SR, non-prompt estimate in TT enriched SR. The nominal configuration of the matrix method is to match the leading lepton with trigger objects. Matching the 2nd-leading 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,
- MET>85GeV,
- Veto events with at least one b-tag jet ($p_T > 20$ 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:

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

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

$$\Delta = |1 - \epsilon| \tag{17.4}$$

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

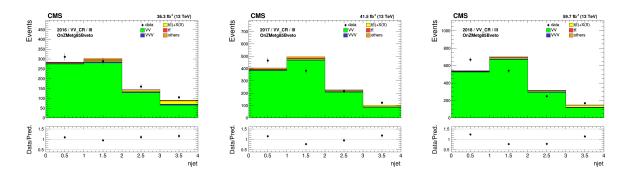


Figure 17.5: Diboson control region, from left to right: 2016, 2017 and 2018 datasets.

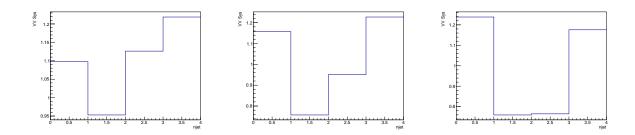


Figure 17.6: Scale factors derived from diboson control region, from left to right: 2016, 2017 and 2018 datasets.

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% (200 GeV-1000 GeV) and is caped at 10% after 1000 GeV.
- 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 ($p_T < 200$ Gev), while GEciteGEmethod method is used for high p_T muons ($p_T > 200$ 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 If, hf, hfstats1/2, and Ifstats1/2 uncertainties. For c jets, we applied cferr1/2 uncertainties. Correlations between different sources are specified in the Table ??.

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.

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.

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
AbsoluteStat	-
AbsoluteScale	✓
AbsoluteMPFBias	✓
Fragmentation	✓
SinglePionECAL	\checkmark
SinglePionHCAL	✓
FlavorQCD	✓
TimePtEta	-
RelativeJEREC1	-
RelativeJEREC2	-
RelativeJERHF	✓
RelativePtBB	✓
RelativePtEC1	-
RelativePtEC2	-
RelativePtHF	✓
RelativeBal	✓
RelativeSample	-
RelativeFSR	-
RelativeStatFSR	✓
RelativeStatEC	-
RelativeStatHF	-
${\sf PileUpDataMC}$	✓
PileUpPtRef	✓
PileUpPtBB	✓
PileUpPtEC1	✓
PileUpPtEC2	✓
PileUpPtHF	✓

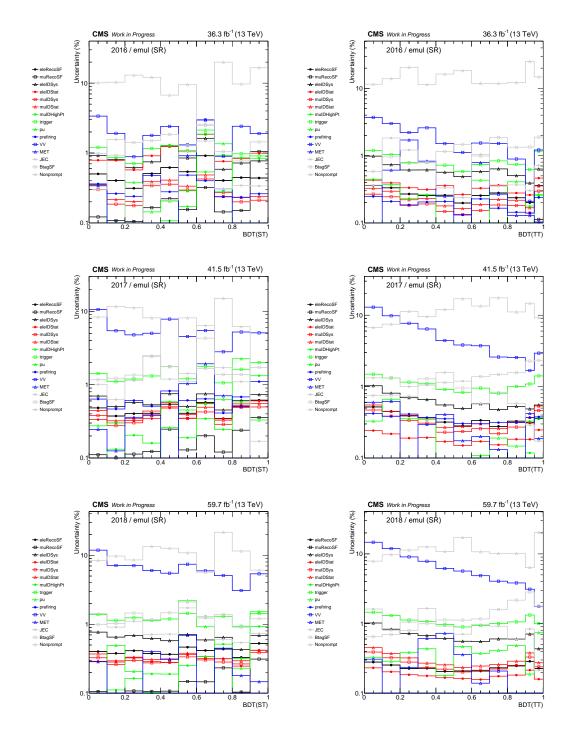


Figure 17.7: Distributions of relative uncertainties on total expected backgrounds as a function of BDT output in ST enriched SR (left), TT enriched SR (right). Luminosity and normalization uncertainties are not included in these plots. JES, JER and HEM are combined into "JEC". Sources of b-tagging uncertainties listed in Table **??** are combined into "BtagSF". From top to bottom: 2016, 2017 and 2018 datasets.

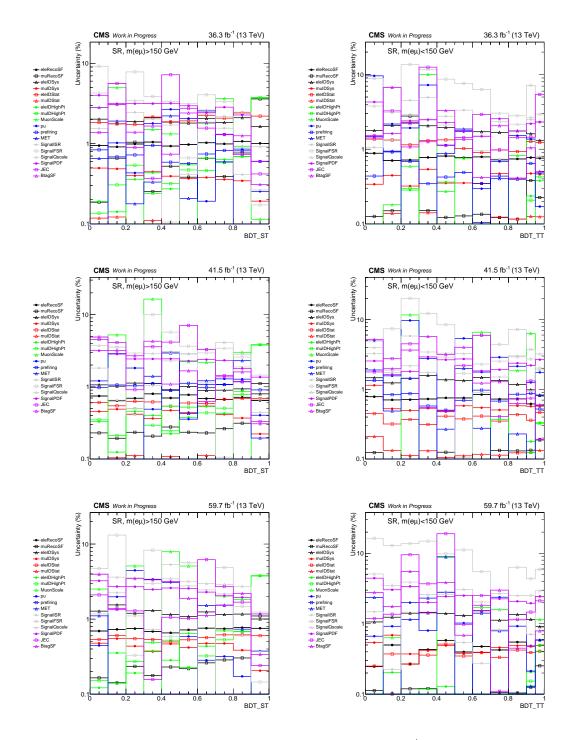


Figure 17.8: Distributions of relative uncertainties on signal ($C_{e\mu tu}^{vector}$ is used as an example) as a function of BDT output in ST enriched SR (left), TT enriched SR (right). Luminosity and normalization uncertainties are not included in these plots. JES, JER and HEM are combined into "JEC". Sources of b-tagging uncertainties listed in Table ?? are combined into "BtagSF". From top to bottom: 2016, 2017 and 2018 datasets.



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. 18.1-18.6. 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 $\mathcal{B}(t\to e\mu c)$ and $\mathcal{B}(t\to e\mu c)$ in the case of nonvanishing signals, the two-dimensional limits can also be obtained through interpolation (see Fig. 18.7). This analysis constitutes the most stringent limits on these processes to date.

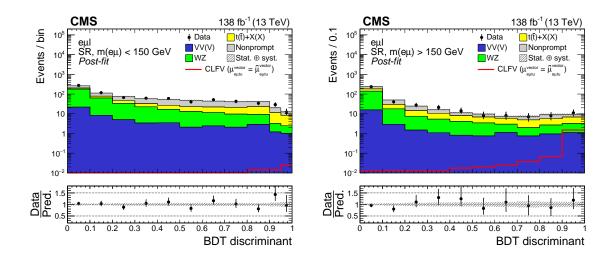


Figure 18.1: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with vector-like operator involving an Up quark. The three data-taking years are aggregated for better visualization.

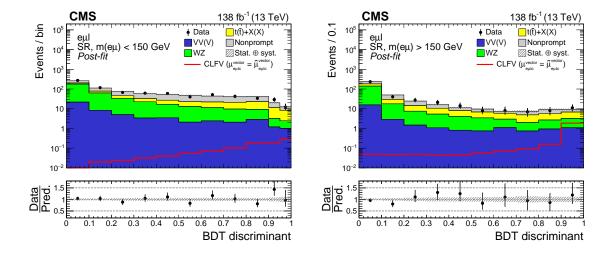


Figure 18.2: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with vector-like operator involving a Charm quark.

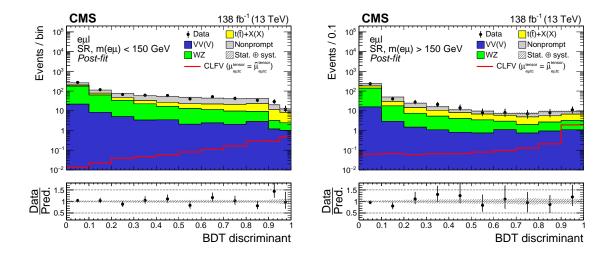


Figure 18.3: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with tensor-like operator involving an Up quark.

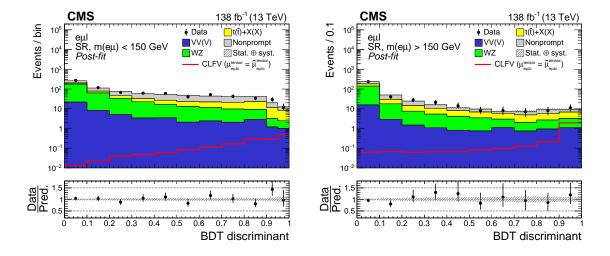


Figure 18.4: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with tensor-like operator involving a Charm quark.

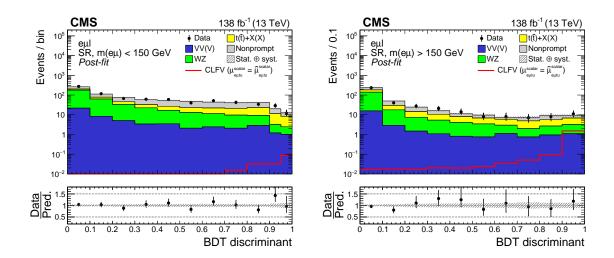


Figure 18.5: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with scalar-like operator involving an Up quark.

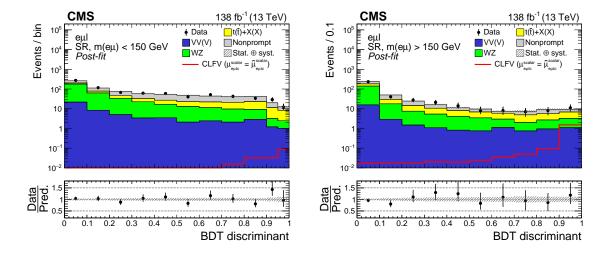


Figure 18.6: Distributions of the posterior BDT discriminator distributions for the TT-enriched SR (left) and the ST-enriched SR (right). Signals are generated with scalar-like operator involving a Charm quark.

CLFV	Lorentz	${\rm C_{e\mu tq}/\Lambda^2~(TeV^{-2})}$		$\mathcal{B}(t o e\muq) imes 10^{-6}$	
coupling	structure	exp (68% range)	obs	exp (68% range)	obs
eµtu	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
eµtc	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.

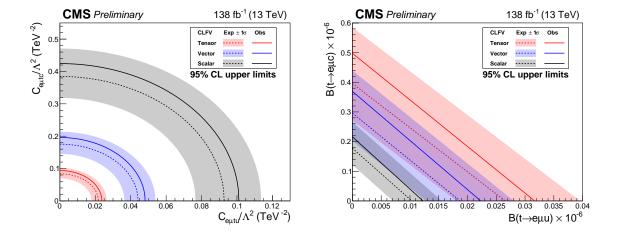


Figure 18.7: Two-dimensional upper limits on the Wilson coefficients (left) and the branching ratios (right).

Part IV

Outlook for CLFV Searches Using Top Quarks

Inclusive Signal Generation

- 19.1 UFO Model
- 19.2 EFT Reweighting

Object and Event Selection

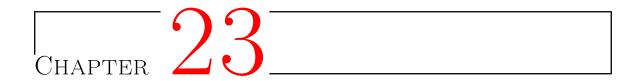
- 20.1 Object Selection
- 20.2 Event Selection

Fake Tau Estimate

- 21.1 ABCD Method
- 21.2 Validation

Expected Sensitivity

- 22.1 Asimov Fit
- 22.2 Expected Upper Limits



Run-2 Combination

- 23.1 Correlation Scheme
- 23.2 Minimal Flavor Violation Interpretation

Conclusion

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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		
	HLT_lsoMu22_eta2p1, HLT_lsoTkMu22_eta2p1		
SingleMuon	HLT_IsoMu24, HLT_IsoTkMu24		
	HLT_Mu50, HLT_TkMu50, HLT_Mu45_eta2p1		
	HLT_Ele25_eta2p1_WPTight_Gsf		
	HLT_Ele27_WPTight_Gsf		
SingleElectron	$HLT_Ele27_eta2p1_WPTight_Gsf$		
SingleElectron	$HLT_Ele32_eta2p1_WPTight_Gsf$		
	HLT_Ele105_CaloIdVT_GsfTrkIdT		
	HLT_Ele115_CaloldVT_GsfTrkIdT		
	HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL		
	HLT_Mu17_TrklsoVVL_TkMu8_TrklsoVVL		
	HLT_TkMu17_TrklsoVVL_TkMu8_TrklsoVVL		
DoubleMuon	HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ		
	HLT_Mu17_TrklsoVVL_TkMu8_TrklsoVVL_DZ		
	HLT_TkMu17_TrklsoVVL_TkMu8_TrklsoVVL_DZ		
	HLT_Mu30_TkMu11, HLT_TripleMu_12_10_5		
	HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL_DZ		
	HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL		
DoubleEG	HLT_DoubleEle33_CaloIdL_MW		
	$HLT_Double Ele 33_Calold L_GsfTrkIdVL$		
	$HLT_Double Ele 33_Calold L_GsfTrkld VL_MW$		
	HLT_Mu23_TrklsoVVL_Ele8_CaloIdL_TrackIdL_IsoVL		
	HLT _Mu23 _TrklsoVVL _Ele8 _CaloIdL _TrackIdL _IsoVL _DZ		
	HLT_Mu8_TrklsoVVL_Ele23_CaloIdL_TrackIdL_IsoVL		
MuonEG	HLT_Mu30_Ele30_CaloldL_GsfTrkldVL		
	HLT _Mu33 _Ele33 _CaloIdL _GsfTrkIdVL		
	HLT_DiMu9_Ele9_CaloIdL_TrackIdL		
	HLT_Mu8_DiEle12_CaloIdL_TrackIdL		

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

Dataset Trigger Path HLT_IsoMu24_eta2p1 HLT_IsoMu27 SingleMuon HLT_Mu50 HLT_OldMu100 HLT_TkMu100 HLT_Ele32_WPTight_Gsf_L1DoubleEG SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkIdT HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8 HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8 HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8 DoubleMuon 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 HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle25_CaloldL_MW HLT_DoubleEle33_CaloldL_MW HLT_DoubleEle33_CaloldL_MW HLT_DiEle27_WPTightCaloOnly_L1DoubleEG	_	
HLT_IsoMu27	Dataset	Trigger Path
SingleMuon HLT_Mu50 HLT_TkMu100 HLT_Ele32_WPTight_Gsf_L1DoubleEG SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkldT HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle25_CaloldL_MW HLT_DoubleEle33_CaloldL_MW		HLT_lsoMu24_eta2p1
HLT_OldMu100 HLT_TkMu100 HLT_Ele32_WPTight_Gsf_L1DoubleEG SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkIdT HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8 HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8 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 HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle25_CaloIdL_MW		HLT_IsoMu27
HLT_TkMu100 HLT_Ele32_WPTight_Gsf_L1DoubleEG SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkIdT HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8 HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8 DoubleMuon 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 HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW	SingleMuon	HLT_Mu50
HLT_Ele32_WPTight_Gsf_L1DoubleEG SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkIdT HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass8 HLT_Mu17_TrkIsoVVL_Mu8_TrkIsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrkIsoVVL_Mu9_TrkIsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_12_10_5 HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_OldMu100
SingleElectron HLT_Ele35_WPTight_Gsf HLT_Ele115_CaloldVT_GsfTrkldT HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle33_CaloldL_MW		HLT_TkMu100
HLT_Ele115_CaloldVT_GsfTrkldT HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_Ele32_WPTight_Gsf_L1DoubleEG
HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle33_CaloldL_MW HLT_DoubleEle33_CaloldL_MW	SingleElectron	HLT_Ele35_WPTight_Gsf
HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8 HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle33_CaloldL_MW HLT_DoubleEle33_CaloldL_MW		HLT_Ele115_CaloIdVT_GsfTrkIdT
HLT _Mu17 _TrklsoVVL _Mu8 _TrklsoVVL _DZ _Mass8 HLT _Mu17 _TrklsoVVL _Mu8 _TrklsoVVL _DZ _Mass3p8 DoubleMuon HLT _Mu19 _TrklsoVVL _Mu9 _TrklsoVVL _DZ _Mass3p8 HLT _Mu37 _TkMu27 HLT _TripleMu _12 _10 _5 HLT _TripleMu _10 _5 _5 _DZ HLT _TripleMu _5 _3 _3 _Mass3p8to60 _DZ HLT _Ele23 _Ele12 _CaloldL _TrackIdL _IsoVL HLT _Ele16 _Ele12 _Ele8 _CaloIdL _TrackIdL DoubleEG HLT _DoubleEle25 _CaloIdL _MW HLT _DoubleEle33 _CaloIdL _MW		HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL
HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8 DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle33_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ
DoubleMuon HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8 HLT_Mu37_TkMu27 HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloldL_TrackIdL DoubleEG HLT_DoubleEle25_CaloldL_MW HLT_DoubleEle33_CaloldL_MW		HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass8
HLT _Mu37 _TkMu27 HLT _TripleMu _12 _10 _5 HLT _TripleMu _10 _5 _5 _DZ HLT _TripleMu _5 _3 _3 _Mass3p8to60 _DZ HLT _Ele23 _Ele12 _CaloldL _TrackIdL _IsoVL HLT _Ele16 _Ele12 _Ele8 _CaloIdL _TrackIdL DoubleEG HLT _DoubleEle25 _CaloIdL _MW HLT _DoubleEle33 _CaloIdL _MW		HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8
HLT_TripleMu_12_10_5 HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloldL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW	DoubleMuon	HLT_Mu19_TrklsoVVL_Mu9_TrklsoVVL_DZ_Mass3p8
HLT_TripleMu_10_5_5_DZ HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_Mu37_TkMu27
HLT_TripleMu_5_3_3_Mass3p8to60_DZ HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_TripleMu_12_10_5
HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_TripleMu_10_5_5_DZ
HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_TripleMu_5_3_3_Mass3p8to60_DZ
DoubleEG HLT_DoubleEle25_CaloIdL_MW HLT_DoubleEle33_CaloIdL_MW		HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL
HLT_DoubleEle33_CaloIdL_MW		HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL
	DoubleEG	$HLT_DoubleEle25_CaloldL_MW$
HLT DiEle27 WPTightCaloOnly L1DoubleEG		HLT_DoubleEle33_CaloIdL_MW
		HLT_DiEle27_WPTightCaloOnly_L1DoubleEG
HLT _Mu23 _ TrklsoVVL _ Ele12 _ CaloIdL _ TrackIdL _ IsoVL		HLT_Mu23_TrklsoVVL_Ele12_CaloIdL_TrackIdL_IsoVL
$HLT_Mu23_TrklsoVVL_Ele12_CaloIdL_TrackIdL_lsoVL_Discrete and the property of the property of$		${\sf HLT_Mu23_TrklsoVVL_Ele12_CaloldL_TrackldL_lsoVL_DZ}$
$HLT_Mu8_TrklsoVVL_Ele23_CaloldL_TrackIdL_IsoVL_DZ$		HLT_Mu8_TrkIsoVVL_Ele23_CaloIdL_TrackIdL_IsoVL_DZ
HLT_Mu12_TrklsoVVL_Ele23_CaloIdL_TrackIdL_IsoVL_D		${\sf HLT_Mu12_TrklsoVVL_Ele23_CaloldL_TrackldL_lsoVL_DZ}$
MuonEG HLT_Mu27_Ele37_CaloIdL_MV	MuonEG	HLT_Mu27_Ele37_CaloIdL_MV
$HLT_Mu37_Ele27_CaloldL_MV$		HLT_Mu37_Ele27_CaloIdL_MV
$HLT_DiMu9_Ele9_CaloldL_TrackIdL$		$HLT_DiMu9_Ele9_CaloldL_TrackIdL$
$HLT_DiMu9_Ele9_CaloldL_TrackIdL_DZ$		$HLT_DiMu9_Ele9_CaloldL_TrackIdL_DZ$
HLT_Mu8_DiEle12_CaloIdL_TrackIdL		HLT_Mu8_DiEle12_CaloIdL_TrackIdL

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

Dataset	Trigger Path	
	HLT_IsoMu24	
	HLT_IsoMu27	
SingleMuon	HLT_Mu50	
	HLT_OldMu100	
	HLT_TkMu100	
	HLT_Ele32_WPTight_Gsf	
	HLT_Ele115_CaloIdVT_GsfTrkIdT	
EGamma	HLT_Ele23_Ele12_CaloIdL_TrackIdL_IsoVL	
LGamma	HLT_Ele16_Ele12_Ele8_CaloIdL_TrackIdL	
	$HLT_DoubleEle25_CaloIdL_MW$	
	HLT_DiEle27_WPTightCaloOnly_L1DoubleEG	
	HLT_Mu17_TrklsoVVL_Mu8_TrklsoVVL_DZ_Mass3p8	
	HLT_Mu37_TkMu27	
DoubleMuon	HLT_TripleMu_12_10_5	
	HLT_TripleMu_10_5_5_DZ	
	HLT_TripleMu_5_3_3_Mass3p8to60_DZ	
	$HLT_Mu23_TrklsoVVL_Ele12_CaloldL_TrackldL_lsoVL$	
	${\sf HLT_Mu23_TrklsoVVL_Ele12_CaloldL_TrackIdL_IsoVL_DZ}$	
	$HLT_Mu8_TrkIsoVVL_Ele23_CaloIdL_TrackIdL_IsoVL_DZ$	
MuonEG	${\sf HLT_Mu12_TrklsoVVL_Ele23_CaloldL_TrackIdL_lsoVL_DZ}$	
	HLT_Mu27_Ele37_CaloIdL_MV	
	HLT_Mu37_Ele27_CaloIdL_MV	
	HLT_DiMu9_Ele9_CaloIdL_TrackIdL_DZ	
	HLT_Mu8_DiEle12_CaloldL_TrackIdL	



Nuisance Parameter Impact

The observed and expected ($\mu_{exp}=0$) impacts of the nuisance parameters on the profile likelihood fit are shown in Figure B.1-B.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 B.3.

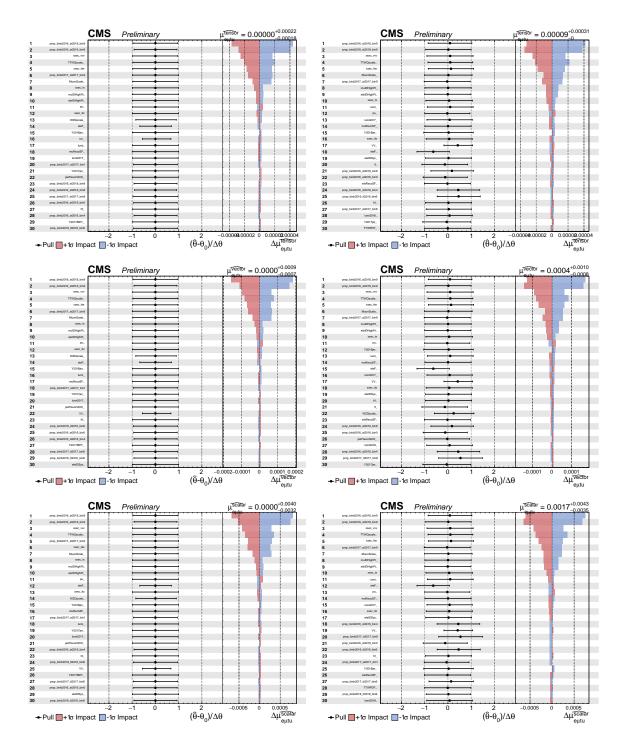


Figure B.1: Impacts of nuissance parameters for run II limit setting. From top to bottom: $e\mu tu$ -tensor, $e\mu tu$ -vector, $e\mu tu$ -scalar. From left to right: expected impact (expected signal strength at 0), observed impact.

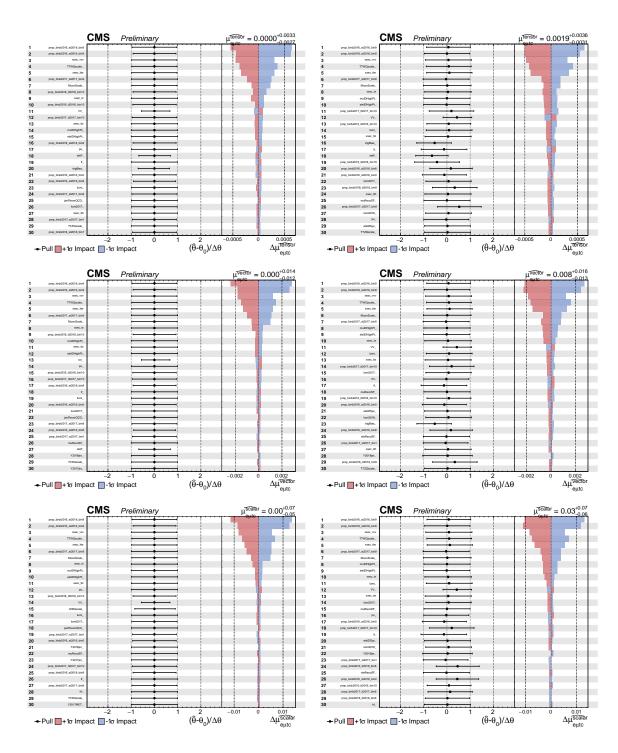


Figure B.2: Impacts of nuissance parameters for run II limit setting. From top to bottom: $e\mu tc$ -tensor, $e\mu tc$ -vector, $e\mu tc$ -scalar. From left to right: expected impact (expected signal strength at 0), observed impact.

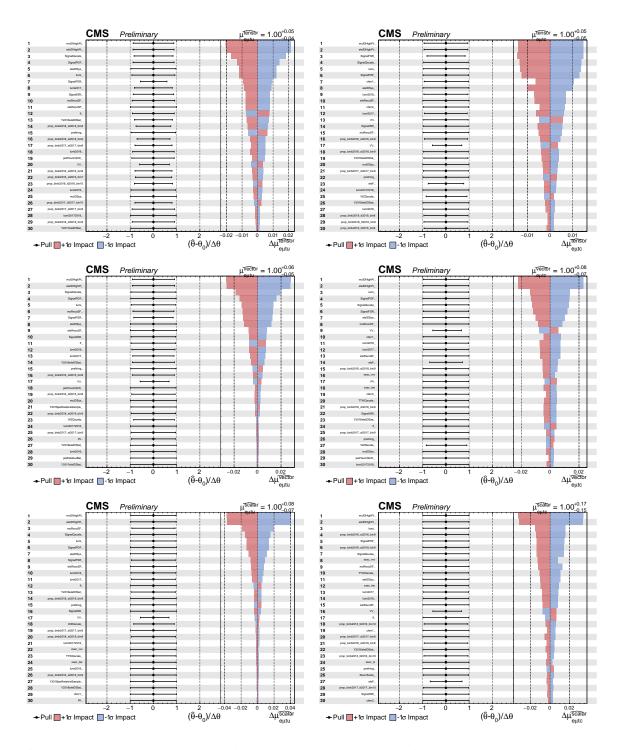


Figure B.3: Expected impact with an expected signal strength at 1. From top to bottom: tensor, vector, scalar. From left to right: $e\mu tu$, $e\mu tc$.



More materials