

Search for Lepton-Flavour-Violating Decays of the Higgs Boson

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on behalf of the ATLAS Collaboration

Higgs 2022

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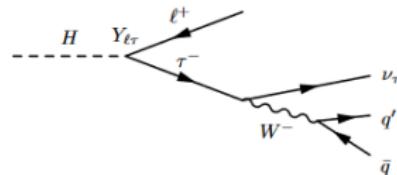
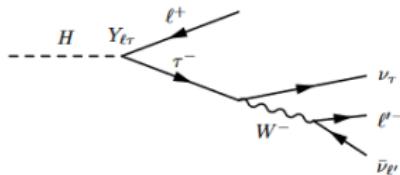


Motivation

- Lepton flavour conserved in the SM.
- Lepton Flavour Violation (LFV) is evident in neutrino oscillations.
Can LFV processes also occur in charged lepton sector?
- Higgs boson LFV decays expected in several SM extensions (SUSY, 2HDM, composite Higgs...).
- This talk is on the results of the ATLAS search for $H \rightarrow e\tau$ and $H \rightarrow \mu\tau$ decays at 138 fb^{-1} . [Link to Conf note](#).

Analysis Introduction

- Search for evidence of Higgs boson decay with charged lepton flavour violation (LFV) with two separate signals, $H \rightarrow e\tau$ and $H \rightarrow \mu\tau$ and two separate analysis methods, **MC-template** and **Symmetry**.
- All major production modes are considered as signal (ggH, VBF, WH, ZH)



- **Lep-Lep** Final states $e\tau_\mu$ and $\mu\tau_e$ with one electron and one muon.
- Ordering based on p_T of leptons in Higgs frame (new to this analysis).
- Two different methods:
 - **MC-template method**
 - **Symmetry method**

- **Lep-Had** Final states $e\tau_{\text{had}}$ and $\mu\tau_{\text{had}}$ with one lepton and one hadronic tau.
- One analysis method:
 - **MC-template method**

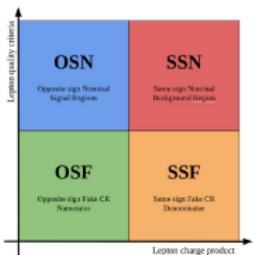
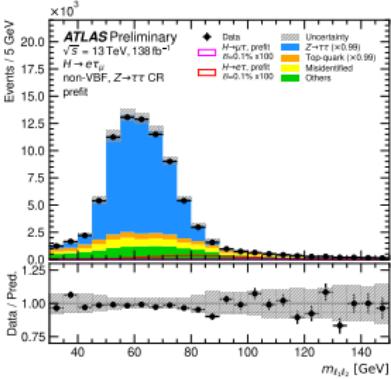
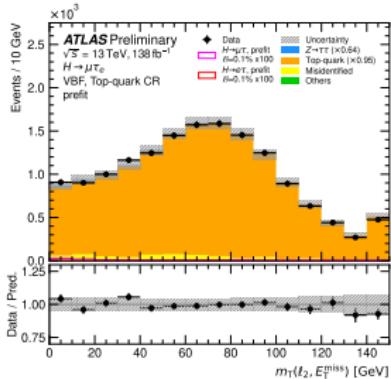
Event selection

Lep-Lep	Lep-Had
Baseline	
1e and 1 μ , Opposite sign	1e or 1 μ and 1tau-had with Opposite sign
No tau-had	
No b-jets	
VBF	
$N_{\text{jets}} \geq 2$	
Non-VBF	
fail VBF selection	

- See Backup for detailed cuts.
- VBF region enhances sensitivity to VBFH production mode using jet kinematics.
- 2 Final states \times 2 regions = 4 SRs per analysis method.
- Additional CRs dependent on the analysis (see next slides).

Background Estimation - Lep-Lep MC Template Method

- **Top** and $Z \rightarrow \tau\tau$: Normalisation factors (NF) extracted separately for VBF and non-VBF from 1-bin CRs.
 - CRs included in fit to constrain background yields.
 - 2-POI CRs shared for $e\tau$ and $\mu\tau$.
 - $Z \rightarrow \mu\mu$: NF from CR at pre-fit level together with norm. unc.
 - **Diboson**: Dedicated Validation region.
 - **SM $H \rightarrow \tau\tau$** and $H \rightarrow WW$ from MC.
 - **Fakes**: Data-driven estimate for $j \rightarrow \ell$, $\gamma \rightarrow e$, $\tau_{\text{had}} \rightarrow \ell$.



Background Estimation - Lep-Had MC Template Method

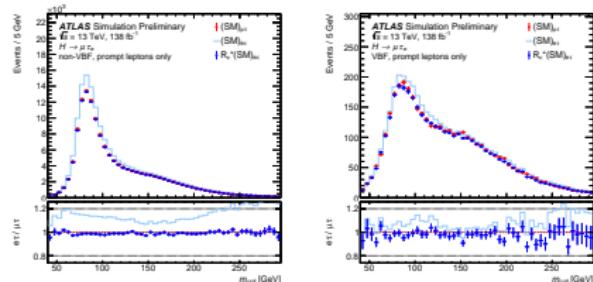
- $Z \rightarrow \tau\tau$: different NF for VBF and non-VBF in the fit.
- **Top**: NF shared with lep-lep for categories where the MC-template is used for lep-lep. Otherwise Top normalization fixed from MC and corresponding theory uncertainties included.
- In 2-POI fit NFs are common for $e\tau_{\text{had}}$ or $\mu\tau_{\text{had}}$.
- $Z \rightarrow \mu\mu$: uncertainties on the normalisation are extracted from dedicated VR.
- **Diboson, SM Higgs, others**: estimated from MC.
- **Fakes**: estimate of $j \rightarrow \tau_{\text{had}}$ with fake-factor. Main sources are $W+\text{jets}$ and multijets.

Background Estimation - Symmetry Lep-Lep Method

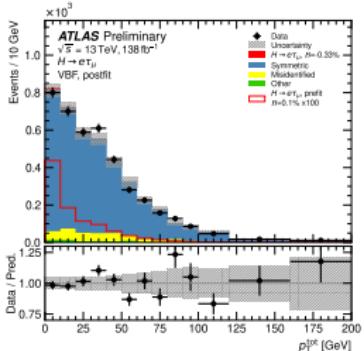
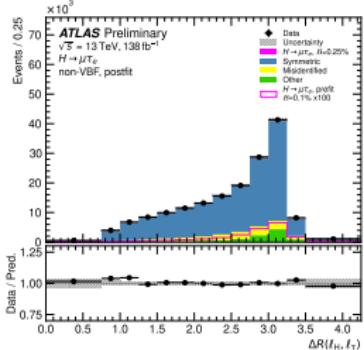
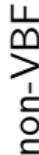
- Data-driven method where main backgrounds in each channel are estimated using data yields from the other channel.
 - SM processes are symmetric w.r.t. prompt $e \leftrightarrow \mu$ exchange.
 - LFV H decays break this symmetry.
 - Data of each of the two channels ($e\tau_\mu$ and $\mu\tau_e$) can serve as background prediction for the other channel.
- The Symmetry method measures the difference of LFV signal strengths: $(\mathcal{B}(H \rightarrow \mu\tau) - \mathcal{B}(H \rightarrow e\tau))$. If one of the signal is assumed to be zero, then it becomes an absolute measurement.

Detector effects induce asymmetries, contribute differently to each channel:

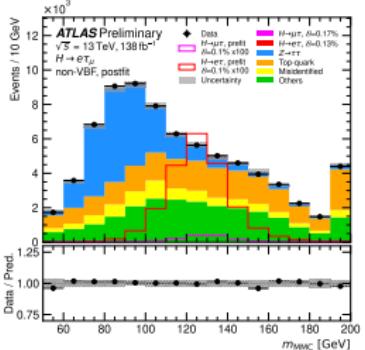
- Misidentified background events
- Different efficiencies for e and μ in reconstruction, identification...



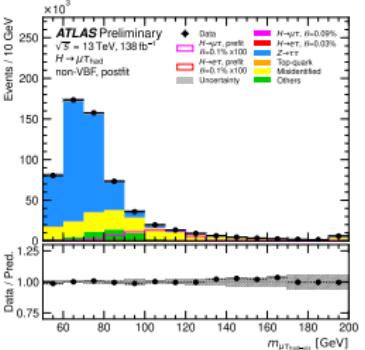
Signal Region



Symmetry Lep-Lep



MC Lep-Lep



MC Lep-Had

MVA Strategy

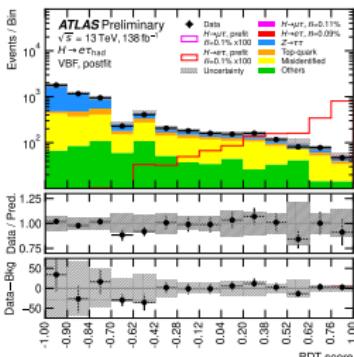
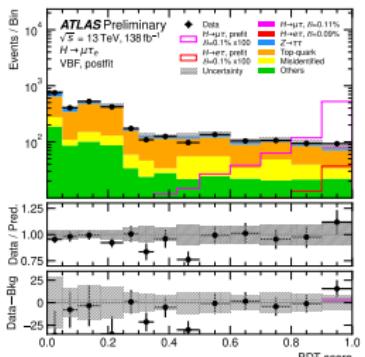
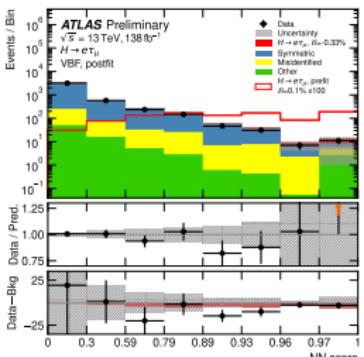
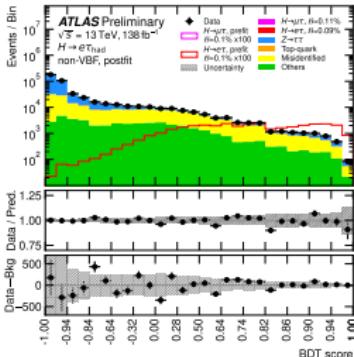
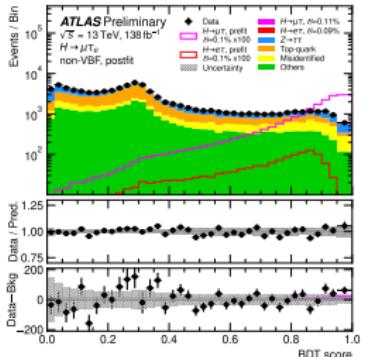
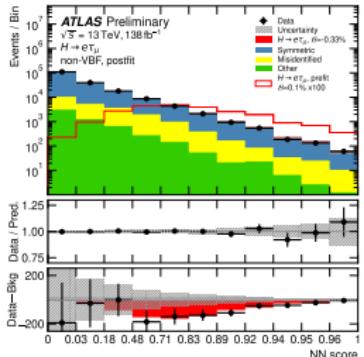
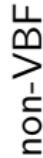
- MVAs used to enhance sensitivity. Scores used as final discriminant.
- Symmetry Lep-Lep: NNs trained with Keras.
- MC-Template: BDTs trained with TMVA.
- Separate training for non-VBF and VBF and for $e\tau_{\text{had}}$ and $\mu\tau_{\text{had}}$ but combined across $e\tau_\mu$, $\mu\tau_e$.

In general: Each MVA linear combination of individual MVAs trained on selected backgrounds.

Example (MC Lep-Lep):

- LFV vs $Z\tau\tau + H\tau\tau + Z\ell\ell$
- LFV vs Top + $VV + HW$
- LFV vs Fakes

Selection of MVA Distributions



Symmetry Lep-Lep

MC Lep-Lep

MC Lep-Had

Statistical Analysis Overview

- MVA output used as final discriminant for signal strength μ extraction.
- Two different signal parametrisations:

Method	Channel	Category	Region	1 POI fit	2 POI fit
MC-template	$\ell\tau_{\ell'}$	non-VBF	SR	✓	✓
			$Z \rightarrow \tau\tau$ CR	✓	✓
			Top-quark CR	✓	✓
	$\ell\tau_{\text{had}}$	VBF	SR	✓	
			$Z \rightarrow \tau\tau$ CR	✓	
			Top-quark CR	✓	
MC-template	$\ell\tau_{\text{had}}$	non-VBF	SR	✓	✓
		VBF	SR	✓	✓
Symmetry	$\ell\tau_{\ell'}$	non-VBF	SR		
			SR	✓	

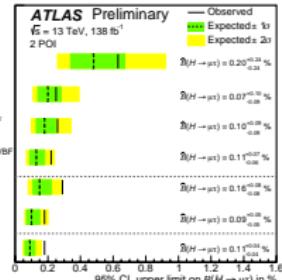
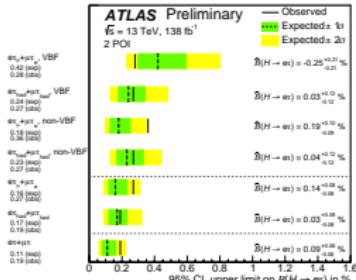
1-POI

- Independent fits in $e\tau$ and $\mu\tau$ channels (for example set $\mathcal{B}(H \rightarrow e\tau) = 0$ when measuring $\mathcal{B}(H \rightarrow \mu\tau)$)
- non-VBF: MC-LepLep + MC-LepHad
- VBF: Symmetry-LepLep + MC-LepHad

2-POI

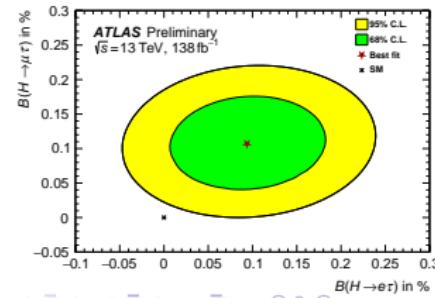
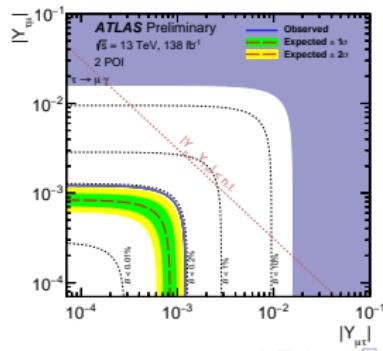
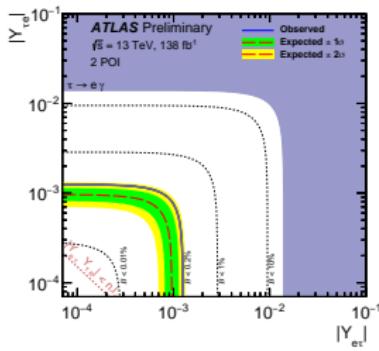
- No assumption needed on branching ratios.
- The two signals are fitted simultaneously in the two channels
- non-VBF + VBF: MC-LepLep + MC-LepHad

2-POI results



$\mathcal{B}(H \rightarrow e\tau)$

$\mathcal{B}(H \rightarrow \mu\tau)$



- Observed (expected) limits at 95% CL are $0.19(0.11)\%$ and $0.18(0.09)\%$ for $\mathcal{B}(H \rightarrow e\tau)$ and $\mathcal{B}(H \rightarrow \mu\tau)$ respectively
- $\Rightarrow 1.6\sigma$ excess for $\mathcal{B}(H \rightarrow e\tau)$ and 2.5σ for $\mathcal{B}(H \rightarrow \mu\tau)$
- \Rightarrow Compatible with SM branching value of 0 to 2.2σ

Conclusion

- Presented the results of a search for $H \rightarrow e\tau$ and $H \rightarrow \mu\tau$ decays with the ATLAS detector at 138 fb^{-1}
- Two categories, Lep-Lep and Lep-had and two regions, VBF-enhanced and non-VBF
- Two different analysis methods: MC-Template and Symmetry(Lep-Lep only)
- Results quoted as 1-POI fits for independent $\mathcal{B}(H \rightarrow e\tau)$ and $\mathcal{B}(H \rightarrow \mu\tau)$ measurements and combined 2-POI measurement.
- Comparison to previous 36 fb^{-1} analysis:
 - **Observed Limits** improved by a factor of 2.4(1.5) for $H \rightarrow e\tau(H \rightarrow \mu\tau)$
 - **Expected Limits** improved by a factor of 3.1(4.1) for $H \rightarrow e\tau(H \rightarrow \mu\tau)$

Backup

Detailed Event selection

Selection	$\ell\tau_{\ell'}$	$\ell\tau_{\text{had}}$
	exactly 1e and 1 μ , OS $\tau_{\text{had-veto}}$	exactly 1 ℓ and 1 $\tau_{\text{had-vis}}$, OS $\tau_{\text{had}} \text{Tight ID}$
<i>Baseline</i>	<i>b</i> -veto $p_T^{\ell_1} > 45$ (35) GeV MC-template (Symmetry method) $p_T^{\ell_2} > 15$ GeV $30 \text{ GeV} < m_{\ell_1\ell_2} < 150 \text{ GeV}$ $0.2 < p_T^{\text{track}}(\ell_2 = e) / p_T^{\text{cluster}}(\ell_2 = e) < 1.25$ (MC-template) track d_0 significance requirement (see text) $ z_0 \sin \theta < 0.5 \text{ mm}$	Medium eBDT ($e\tau_{\text{had}}$) <i>b</i> -veto $p_T^\ell > 27.3$ GeV $p_T^{\tau_{\text{had-vis}}} > 25 \text{ GeV}, \eta^{\tau_{\text{had-vis}}} < 2.4$ $\sum_{i=\ell, \tau_{\text{had-vis}}} \cos \Delta\phi(i, E_T^{\text{miss}}) > -0.35$ $ \Delta\eta(\ell, \tau_{\text{had-vis}}) < 2$
<i>VBF</i>	<i>Baseline</i>	
	≥ 2 jets, $p_T^{j_1} > 40$ GeV, $p_T^{j_2} > 30$ GeV $ \Delta\eta_{jj} > 3, m_{jj} > 400$ GeV	
<i>non-VBF</i>	<i>Baseline</i> plus fail VBF categorisation	
	—	veto events if $90 < m_{\text{vis}}(e, \tau_{\text{had-vis}}) < 100$ GeV

Uncertainty Sources

1 POI Source of uncertainty	Impact ($\times 10^2$) on observed $\hat{B}(H \rightarrow e\tau)$ $\hat{B}(H \rightarrow \mu\tau)$	
Flavour tagging	0.010	0.003
Misidentified background ($e\tau_{\text{had}}$)	0.021	0.015
Misidentified background ($\ell\tau_{\ell'}$)	0.029	0.016
Jet and E_T^{miss}	0.011	0.010
Electrons and muons	0.003	0.005
Luminosity	0.006	0.006
Hadronic τ decays	0.009	0.009
Theory (signal)	0.008	0.006
Theory ($Z + \text{jets}$ processes)	0.009	0.011
Theory (top-quark processes)	0.003	0.003
$Z \rightarrow \ell\ell$ normalisation	0.002	0.006
Symmetric background estimate	0.002	0.001
Background sample size	0.042	0.023
Total systematic uncertainty	0.053	0.038
Data sample size	0.030	0.028
Total	0.061	0.047

2 POI Source of uncertainty	Impact ($\times 10^2$) on observed $\hat{B}(H \rightarrow e\tau)$ $\hat{B}(H \rightarrow \mu\tau)$	
Flavour tagging	0.007	0.003
Misidentified background ($e\tau_{\text{had}}$)	0.021	0.003
Misidentified background ($e\tau_\mu$)	0.058	0.003
Misidentified background ($\mu\tau_{\text{had}}$)	0.006	0.015
Misidentified background ($\mu\tau_e$)	0.009	0.011
Jet and E_T^{miss}	0.012	0.009
Electrons and muons	0.013	0.005
Luminosity	0.007	0.005
Hadronic τ decays	0.009	0.009
Theory (signal)	0.007	0.007
Theory ($Z + \text{jets}$ processes)	0.007	0.009
$Z \rightarrow \ell\ell$ normalisation ($e\tau$)	< 0.001	< 0.001
$Z \rightarrow \ell\ell$ normalisation ($\mu\tau$)	0.002	0.007
Background sample size	0.037	0.023
Total systematic uncertainty	0.051	0.036
Data sample size	0.030	0.027
Total	0.059	0.045

MVA Input Variables

Variable	$\ell\tau_{\text{had}}$		$\ell\tau_{\ell'}$ MC-template		$\ell\tau_{\ell'}$ Symmetry	
	non-VBF	VBF	non-VBF	VBF	non-VBF	VBF
m_{coll}	✓	✓	✓	✓	✓	✓
m_{vis}	✓	✓	✓	✓	✓	✓
m_{MMC}			✓	✓	✓	✓
$m_{\text{T}}(\tau, E_{\text{T}}^{\text{miss}})$	✓	✓			✓	✓
$m_{\text{T}}(\ell_H, E_{\text{T}}^{\text{miss}})$	✓	✓			✓	✓
$m_{\text{T}}(\ell_1, E_{\text{T}}^{\text{miss}})$			✓	✓		
$m_{\text{T}}(\ell_2, E_{\text{T}}^{\text{miss}})$			✓	✓		
$E_{\text{T}}^{\text{miss}}$	✓	✓	✓	✓	✓	✓
$p_{\text{T}}(\ell_H)$	✓	✓				
$p_{\text{T}}(\tau)$	✓	✓				
$p_{\text{T}}(\ell_{\tau})/p_{\text{T}}(\ell_1)$			✓	✓		
$p_{\text{T}}^{\text{rest}}(\ell_H)$			✓	✓	✓	✓
$p_{\text{T}}^{\text{rest}}(\ell_{\tau})$			✓	✓	✓	✓
$p_{\text{T}}^{\text{tot}}$				✓	✓	✓
$p_{\text{T}}(\ell_H)/E_{\text{T}}^{\text{miss}}$					✓	✓
$p_{\text{T}}(\tau)/p_{\text{T}}(\ell_H)$					✓	✓
$\sum p_{\text{T}}$					✓	✓
$\Delta R(\ell_H, \tau)$	✓	✓	✓	✓	✓	✓
$\Delta\eta(\ell_H, \tau)$	✓	✓			✓	✓
$\Delta\phi(\ell_H, \tau)$	✓				✓	
$\Delta\phi(\ell_{\tau}, E_{\text{T}}^{\text{miss}})$			✓	✓	✓	✓
$ \Delta\phi(\ell_H, E_{\text{T}}^{\text{miss}}) - \Delta\phi(\tau, E_{\text{T}}^{\text{miss}}) $	✓					
$\Delta\alpha$			✓	✓	✓	✓
$\Delta\Phi(\ell_H, E_{\text{T}}^{\text{miss}})$			✓	✓	✓	✓
$\Delta d_0(\ell_1, \ell_2)$			✓	✓	✓	
$\sigma_{d_0}^{\ell_{\tau}}$			✓	✓		
$\eta(\tau_{\text{had-vis}})$	✓	✓				
m_{ij}		✓				✓
$N_{\text{jets}}(p_{\text{T}} > 30\text{GeV})$		✓				
$ \Delta\eta_{jj} $		✓				
$\Delta R(j, j)$					✓	✓
$ \Delta\eta_{jj} \cdot \eta_{j_1} \cdot \eta_{j_2} $			✓			
$p_{\text{T}}(j_1)$						✓
$p_{\text{T}}(j_2)$						✓
$\Delta\phi(j_1, E_{\text{T}}^{\text{miss}})$				✓		✓
$\Delta\phi(j_2, E_{\text{T}}^{\text{miss}})$				✓		✓
$\eta\text{-centrality}(\ell_H)$			✓		✓	✓
$\eta\text{-centrality}(\ell_{\tau})$			✓		✓	



MVA Parameters

Region	Channel	NTrees	MaxDepth	MinNodeSize	Shrinkage
<i>non-VBF</i>	$e\tau_{\text{had}}, \mu\tau_{\text{had}}$	500	7	1	0.05
<i>VBF</i>	$e\tau_{\text{had}}$	300	10	1	0.01
	$\mu\tau_{\text{had}} \text{ BDT}_1$	300	8	1	0.009
	$\mu\tau_{\text{had}} \text{ BDT}_2$	300	6	1	0.0095
<i>non-VBF, VBF</i>	$\ell\tau_{\ell'}$	750	8	2.5	0.1

Hyperparameter	Value			
	<i>non-VBF NN</i>	<i>VBF$Z \rightarrow \tau\tau$ NN</i>	<i>VBF$_{\text{top-quark}}$ NN</i>	<i>VBF$_{\text{misID}}$ NN</i>
# nodes in 1st layer	512	128	128	128
# hidden layers	2	4	3	4
# output layers	3	2	2	2
L2 weight reg. param.	0.000048	0.000292	0.000094	0.000356
leaky ReLU slope below 0	0.080537	0.019614	0.062515	0.084219
optimiser	SGD	Adam	Adam	Adam
learning rate	0.025810	0.000142	0.000215	0.003507
batch size	128	128	512	1024
epochs	100	100	100	100

MVA Strategy

Symmetry Lep-Lep

- NNs trained with Keras
- Separate training for non-VBF and VBF, but common for $e\tau_\mu$ and $\mu\tau_e$

MC-Template Lep-Lep

- BDTs with TMVA
- Separate training for non-VBF and VBF but common for $e\tau_\mu$ and $\mu\tau_e$

MC-Template Lep-Had

- BDTs with TMVA
- Separate training for non-VBF and VBF and for $e\tau_{had}$ and $\mu\tau_{had}$

non-VBF

1 Multiclassifier NN with 3 output nodes. Signal output node used for fit.

non-VBF and VBF

- 3 BDTs, combined linearly.
- LFV vs $Z\tau\tau + H\tau\tau + Z\ell\ell$
 - LFV vs $\text{Top} + VV + HW$
 - LFV vs Fakes

non-VBF $e\tau_{had}$

- 3 BDTs, combined linearly.
- LFV vs $Z\tau\tau$
 - LFV vs Fakes
 - LFV vs Rest of bkg

VBF

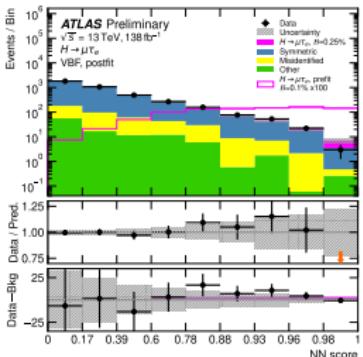
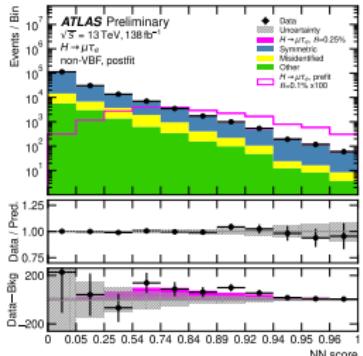
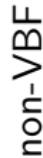
3 MVAs. Scores combined linearly.

- LFV vs. $Z\tau\tau + H\tau\tau + \text{MCfakes}$
- LFV vs. $\text{Top} + VV + HW$
- LFV vs. Fakes

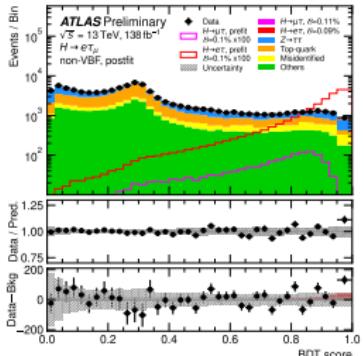
VBF and non-VBF $\mu\tau_{had}$

- 2 BDTs, combined linearly for non-VBF $\mu\tau_{had}$ and quadratically for VBF
- LFV vs $Z\tau\tau$
 - LFV vs Rest of bkg

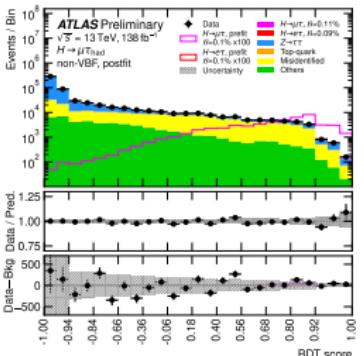
Additional MVA Distributions



Symmetry Lep-Lep

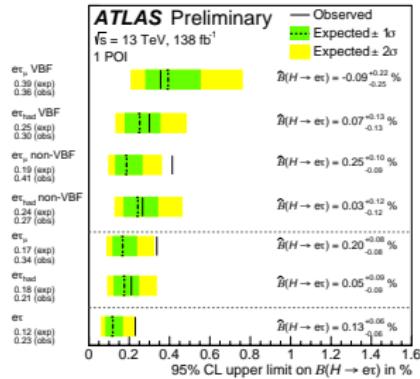


MC Lep-Lep



MC Lep-Had

1-POI results



- Observed (expected) limits at 95% CL are 0.23(0.12)% and 0.16(0.09)% for $\mathcal{B}(H \rightarrow e\tau)$ and $\mathcal{B}(H \rightarrow \mu\tau)$ respectively

