# Search for Flavor Changing Neutral Currents in Top Quark Decays

$$t \rightarrow q \gamma$$

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 $e \rightarrow \gamma$  Fake Rate: Initial Studies

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### The Standard Model

#### **Standard Model of Elementary Particles**

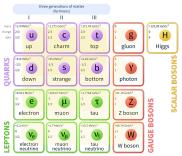


Figure: List of standard model particles

- ▶ Our current theory that attempts to explain everything
  - Experimentally precise and well behaved
  - Very few exceptions (i.e. Neutrino Mass, Matter-Antimatter Asymmetry, Dark Matter Abundance)

### The Top Quark

- ► Heaviest fundamental particle, 172.5 *GeV*
- ► Lifetime  $5x10^{-25}s$ , decays before hadronization
  - Allows us to study the decay of a single quark

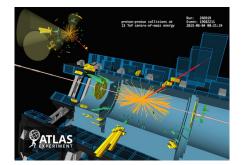


Figure:  $t\bar{t}$  event in the ATLAS detector

### Top Quark Pair Production

- ► Leading order processes for top quark production
  - ▶ Quark-antiquark annihilation  $\approx 10\%$
  - ► Gluon-gluon fusion  $\approx 90\%$

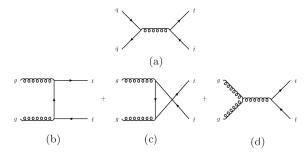


Figure: Leading order  $t\bar{t}$  diagrams

### Top Quark Pair Production

• At  $\sqrt{s} = 13 TeV$  for  $m_t = 172.5 GeV$ ,  $\sigma_{t\bar{t}} = 831.76 pb$ 

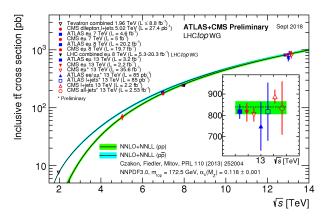


Figure:  $t\bar{t}$  production cross section [TopWGSummaryPlots]

# Top Quark Decays

Standard model top branching ratio to bW  $\simeq 100\%$ 

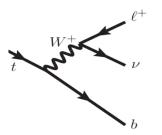


Figure: Leptonic final state diagram for a top decay

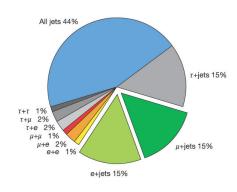
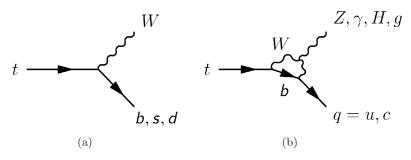


Figure: Top quark pair decay final states [Nature]

### Top Quark Decays in the SM

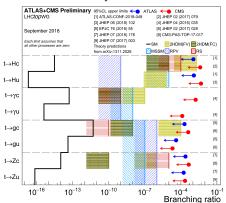


- $ightharpoonup t 
  ightarrow bW \approx 99.83\%$
- ightharpoonup t 
  ightarrow sW pprox 0.16%
- ightharpoonup t 
  ightharpoonup dW pprox 0.01%

• 
$$t \to q_{u,c} X \approx 10^{-17} - 10^{-12}$$

### Top Flavor Changing Neutral Currents

Current Limits on FCNC Decays



- ▶ Limits on  $t \rightarrow \gamma q$  processes: arXiv:1908.08461
  - ►  $t \to \gamma u < 6.1 \times 10^{-5}$
  - ►  $t \to \gamma c < 2.2 \times 10^{-4}$

# Monte Carlo Production of FCNC Signal Samples

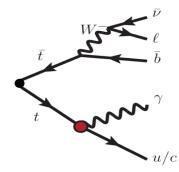
- ► Due to the low cross sections we must create our own Monte Carlo Samples for our Signal
- ► An effective field theory approach was taken in the creation of the model
- ► This model takes advantage of dimension-6 operators

$$\mathcal{L}_{SM} = \mathcal{L}_{SM}^{(4)} + \mathcal{L}^{eff}$$
 where  $\mathcal{L}^{eff} = rac{1}{\Lambda^2} \sum_k C_k^{(6)} Q_k^{(6)}$ 

$$\mathcal{L}_{tq\gamma}^{eff} = C\sigma^{\mu
u}q_{
u}(\lambda_{ct}^{L}P_{L} + \lambda_{ct}^{R}P_{R})tA_{\mu} + H.c.$$

# FCNC: What are we looking for? $t \bar{t} o W( o l u) b + q \gamma$

- ► Final state topology
  - One Neutrino, from W
  - ► One Lepton, from W
  - One B-jet, SM top
  - One Photon, FCNC Top
  - ► One Jet, FCNC Top

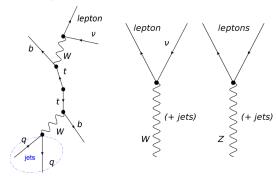


### Object Preselection

- We preselect events with objects that look like our expected topology
- ► Require:
  - **Exactly one lepton (e or**  $\mu$ )  $\geq$  25 GeV
  - ► Exactly one Good photon ≥ 25GeV
  - ► Missing Transverse Energy ≥ 30GeV
  - ► ≥ 2 Jets (at least one being b-tagged)

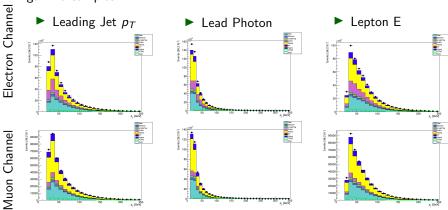
### **Background Processes**

- ▶ Due to all of the processes at hadron colliders it is important to model similar event topologies well
- ▶ Major backgrounds include  $t\bar{t}$ , W+Jets, Z+Jets, + processes with an associated photon



### Preselection Objects

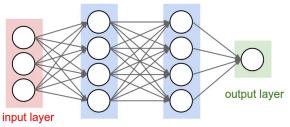
Signal MC Scaled to 1%  $\sigma_{t\bar{t}}$ , No Overlap Removal but no dedicated gamma samples



No Overlap Removal but no dedicated gamma samples - Overall Scale Factors Needed

### Neural Network Architecture

- Using Keras on top of Tensorflow various input parameters are tested for model behavior
- ► A Dense Neural Network with variable number of input variables and hidden layers are explored
- Cut optimization has been performed with full Run 2 luminosity for potential reach of the search



hidden layer 1 hidden layer 2

Figure: [Ref: Neural Network]

# Neural Network Model Inputs

- Using keras on top of tensorflow various input parameters are tested for model behavior
- ▶ Networks are set up with 1 input layer, 2 hidden layers with 10 nodes (+1 bias node) [Ref: Bias], and 1 output node
- ► Each hidden layer has 20% dropout to prevent overtraining by removing codependency between nodes
- ► Batch size of 100 used and each network is allowed 200 epochs (with patience=50), all models converge and end early with reasonable batch sizes
- ► Optimizer: Adam
- ► Loss Function: Binary Cross Entropy

### Cut Optimization

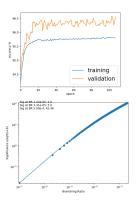
 Reweighting the number of events the model saw by taking advantage of the loss function helps signal/background discrimination

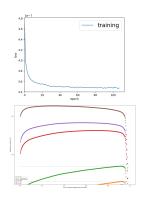
$$\mathsf{Loss} = -\frac{1}{N} \sum_{i=1}^{N} y_i \mathsf{log}(p(y_i)) + (1 - y_i) \mathsf{log}(1 - p(y_i))$$

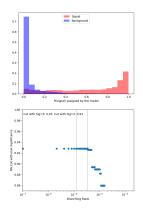
- ▶ y binary indicator (0 or 1) if class label is the correct classification for observation
- ▶ p predicted probability observation is the class label (0 or 1)

### Neural Network Example

### $\mu$ +jets Channel Example

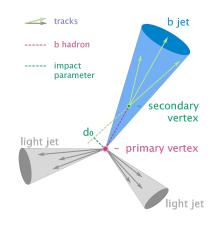




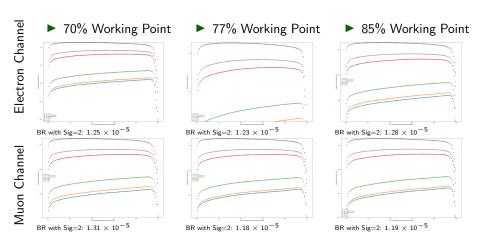


### **B**-tagging

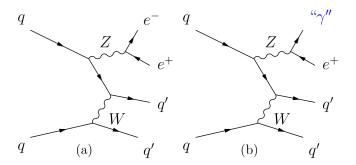
- ► B Hadrons travel a measureable distance before decay
- Tracks originate from outside of interaction point (Seconday Vertex)
- Backtracking tracks in displaced vertex gives an impact parameter
- Decay chain MVA attempts to reconstruct decay of the jet
- Outputs of these algorithms used in a BDT to determine if a Jet is from a b-quark



# Neural Network B-Tagging Results



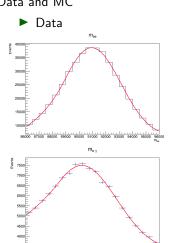
### Fake Rate Studies



Want to be able to correct the number of fake photons predicted in MC to those present in Data



### Data and MC



87000 88000 89000 90000 91000 92000 93000 94000 95000

# Monte Carlo m<sub>e v</sub>

87000 88000 89000 90000 91000 92000 93000 94000 9500

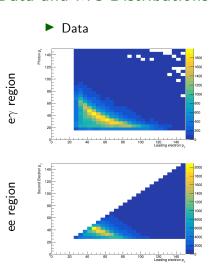
### Scale Factor

$$ext{FR}^{ ext{e-fake}} = rac{ extit{N}_{ ext{e},\gamma}}{ extit{N}_{ ext{e},e} + extit{N}_{ ext{e},\gamma}} \ ext{SF}^{ ext{e-fake}}_{ ext{FR}} = rac{ ext{FR}^{ ext{e-fake}}_{ ext{data}}}{ ext{FR}^{ ext{e-fake}}_{ ext{MC}}}$$

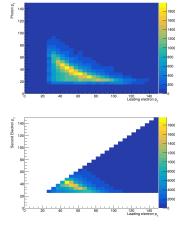
Basic Scale Factor can be calculated for the entire spectrum:

$$\begin{array}{l} \text{FR}_{\text{data}}^{\text{e-fake}} = 0.201 \\ \text{FR}_{\text{MC}}^{\text{e-fake}} = 0.212 \\ \text{SF}_{\text{FR}}^{\text{e-fake}} = 0.953 \end{array}$$

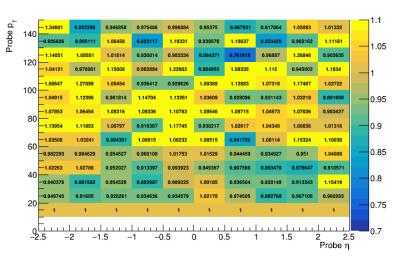
### Data and MC Distributions



# ► Monte Carlo



### 2D Fake Rate



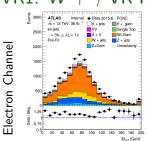
Fake rate will be recalculated using the newest ntuples soon

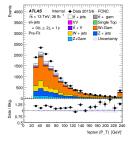
# **Analysis Plots**

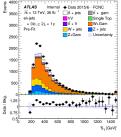
	$VR1:W + \gamma$	$VR2:t\bar{t} + \gamma$	VR3:tt/W+Jets	CR1: W+Jets Rich	CR2: tt̄ Rich	SR
$n_{\gamma}$	=1	=1	=0	= 0	=0	=1
<sup>n</sup> jet	≥2	≥4	=4	=3	≥5	≥2
n <sub>bjet</sub>	=0	=1	=1	=1	=1	=1
<sup>n</sup> lepton	=1	=1	=1	=1	=1	=1
Neural Network Cut	-	<nncut< td=""><td>-</td><td></td><td>-</td><td>&gt;NNCu</td></nncut<>	-		-	>NNCu

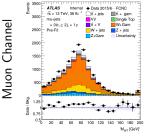
Additional cuts are employed as well to slim out certain backgrounds. The following plots are for MC16a and Data 2015/2016, signal is scaled to 1%  $\sigma_{t\bar{t}}$  except for SR where it is scaled to 0.01% $\sigma_{t\bar{t}}$ 

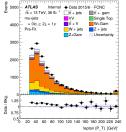
# VR1: $W + \gamma$ VR Plots - $m_T^W$ , lepton $p_T$ , $S_T$

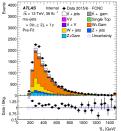


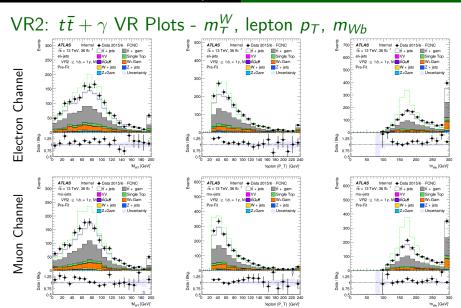


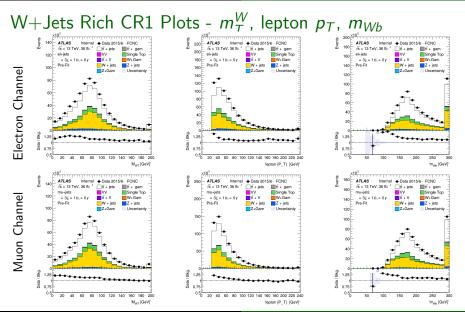


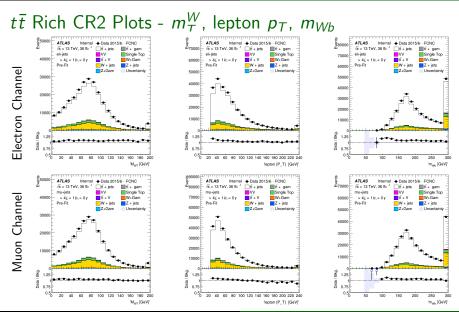




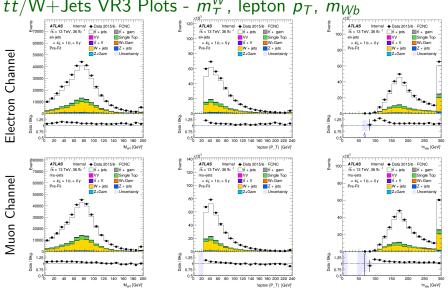


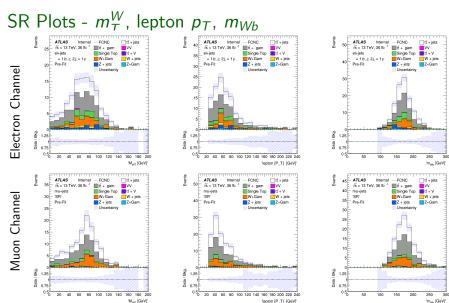






# $t\bar{t}/W+$ Jets VR3 Plots - $m_T^W$ , lepton $p_T$ , $m_{Wb}$



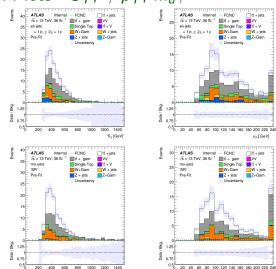


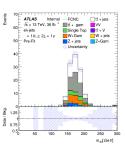
Ye J GeV

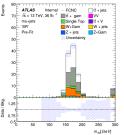
# SR Plots - $S_T$ , $\gamma$ $p_T$ , $m_{a\gamma}$

Electron Channel

Muon Channel



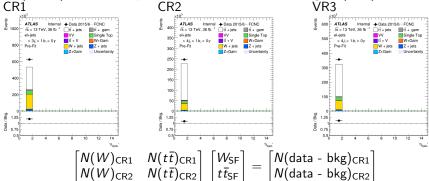




'S, [GeV]

# Scale Factor Calculation based on CR1/CR2

Further scale factors will be applied to the samples to account for mismodeling of the MC. SF calculated based on the control regions (CR1,2) and test its performance in the validation region (VR3)



### To-Do

- ► Checking all MC/Data not just MC16a/2015/16
- ightharpoonup Calculation and verification of Scale Factor for  $t\bar{t}$  and W+Jets
- lacktriangle Recalculation of  $e \to \gamma$  Fake Rate SF using updated samples
- ightharpoonup Calculation of  $j \to \gamma$  Fake Rate SF using Data Driven ABCD method
- Systematics Grid Run on Signal Region to get systematic variations
   Learning how to apply systematics to TRExFitter
- ► Learning how to do *CL<sub>S</sub>* method for limit setting in TRExFitter Framework

### Timeline

- ▶ Dissertation Chapters 1-4 Complete (4 not copyedited yet)
  - ► Chapter 1 Very Brief Introduction
  - ► Chapter 2 Theoretical Background
  - Chapter 3 The LHC and ATLAS
  - ► Chapter 4 Event Simulation and Object Reconstruction
- ► Aiming to continue writing about 1 chapter per month, although it gets a little iffy with the Analysis Chapter
  - ► Remaining Chapters
    - Chapter 5: Search Strategy Background Discussion, NN Discussion
    - Chapter 6: Analysis and Results
    - ► Chapter 7: Outlook for Future Searches
    - Various appendices
- ► Applying to Medical Physics Fellowships over the next 2 months
  - ► All appointments start July 1
- Finishing analysis to-do list by mid-late Jan.
- ▶ Would like to plan defense in mid-late Feb.

# Backup

### Neural Network Model Inputs

Separation =  $\sum_{i}^{bins} \frac{n_{si} - n_{bi}}{n_{si} + n_{bi}}$ 

### mu+jets channel

Variable	Separation
photon0iso	41.18
mqgam	28.27
photon0pt	24.07
mtSM	11.60
mlgam	7.56
deltaRjgam	5.64
deltaRbl	4.42
MWT	3.34
ST	3.30
nuchi2	3.12
jet0pt	2.81
njets	2.07
smchi2	1.89
wchi2	1.87
jet0e	1.52
deltaRlgam	1.17
leptone	0.87
deltaRjb	0.86
met	0.68
bjet0pt	0.52
leptoniso	0.27

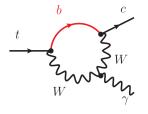
### e+jets channe

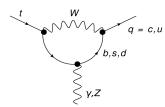
e+jets	Channel
Variable	Separation
photon0pt	23.14
mqgam	22.73
photon0iso	18.70
mtSM	11.02
mlgam	9.53
deltaRbl	5.00
deltaRjgam	4.60
ST	3.83
MWT	3.16
jet0pt	2.47
njets	1.70
nuchi2	1.59
deltaRlgam	1.40
wchi2	1.33
smchi2	1.09
deltaRjb	0.88
leptone	0.85
leptoniso	0.56
bjet0pt	0.50
met	0.47

### Input Variables

```
\label{eq:continuous} \begin{tabular}{ll} ['photon0iso', 'photon0pt', 'mqgam', 'mlgam', 'mtSM', 'deltaRjgam', 'deltaRbl', 'MWT', 'ST', 'njets', 'wchi2', 'jet0pt', 'deltaRlgam', 'leptone', 'met', 'bjet0pt'] \end{tabular}
```

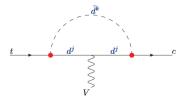
### FCNC Diagrams

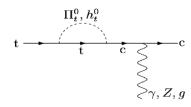




### A Couple of BSM Diagrams

- R-parity-violating supersymmetric models
   [arXiv:hep-ph/9705341]
- Top-color-assisted technicolor models
   [arXiv:hep-ph/0303122]



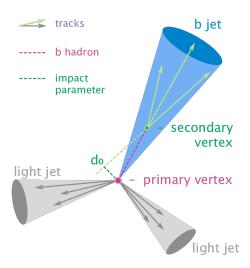


# Jets/AntiKT

$$d_{ij} = min(rac{1}{p_{ti}^2},rac{1}{p_{tj}^2})rac{\Delta_{ij}^2}{R^2}$$
  $d_{iB} = rac{1}{p_{ti}^2}$   $\Delta_{ij}^2 = (\eta_i - \eta_j)^2 + (\phi_i - \phi_j)^2$ 

- ▶ Find minimum of entire set of  $\{d_{ij}, d_{iB}\}$
- ▶ If  $d_{ij}$  is the minimum particles i,j are combined into one particle and removed from the list of particles
- ► If d<sub>iB</sub> is the minimum i is labelled as a final jet and removed from the list of particles
- ▶ Repeat until all particles are part of a jet with distance between jet axes  $\Delta_{ii}$  is greater than R

### **B**-tagging



$$\mathcal{L}_{tq\gamma}^{eff} = -e\bar{c}\frac{i\sigma^{\mu\nu}q_{\nu}}{m_{t}}(\lambda_{ct}^{L}P_{L} + \lambda_{ct}^{R}P_{R})tA_{\mu} + H.c.$$