

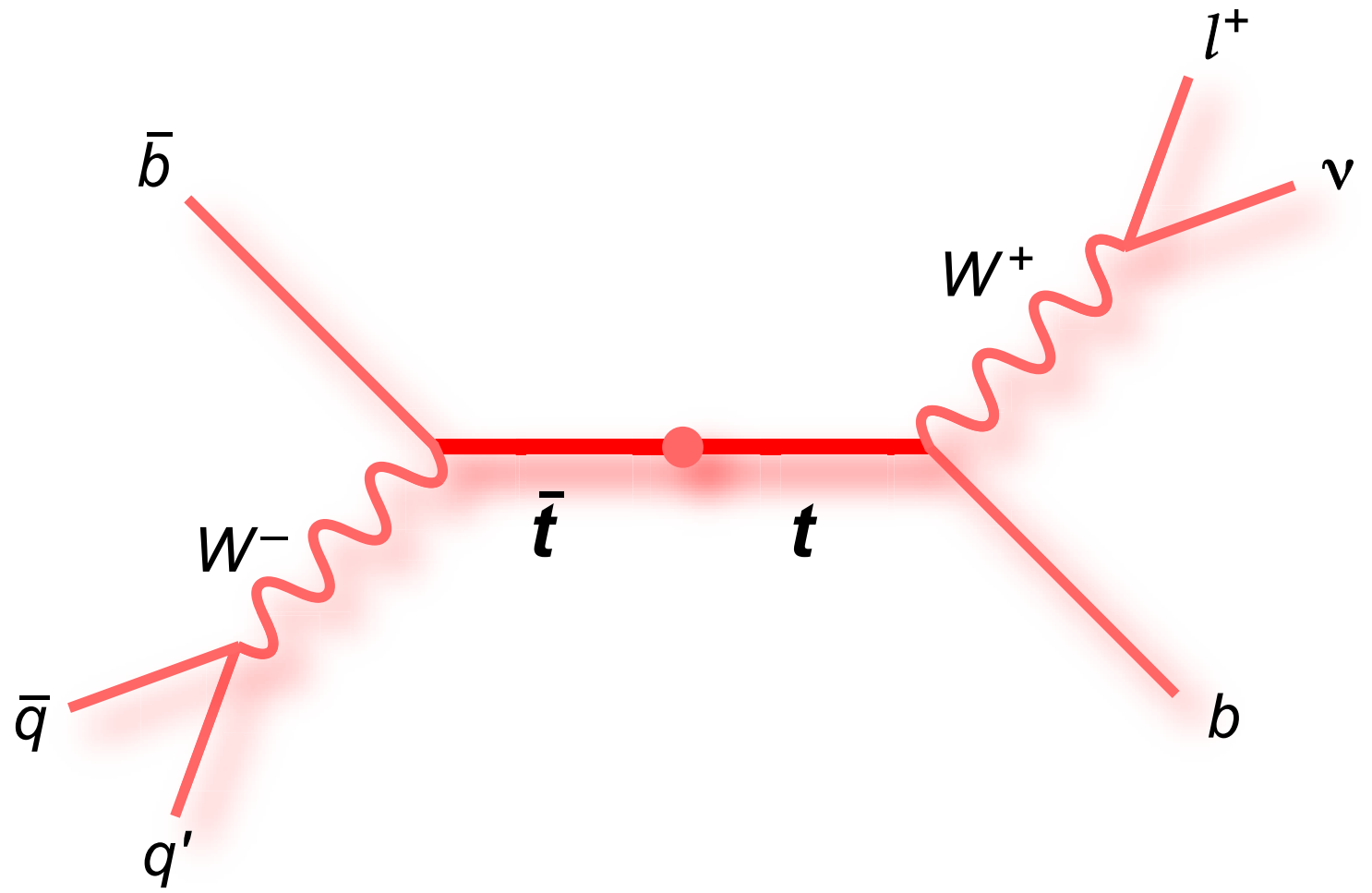
# Reconstrução do ZPrime e Machine Learning utilizando dados de ATLAS

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LIP - UMinho



Decaimento  $T\bar{T}$



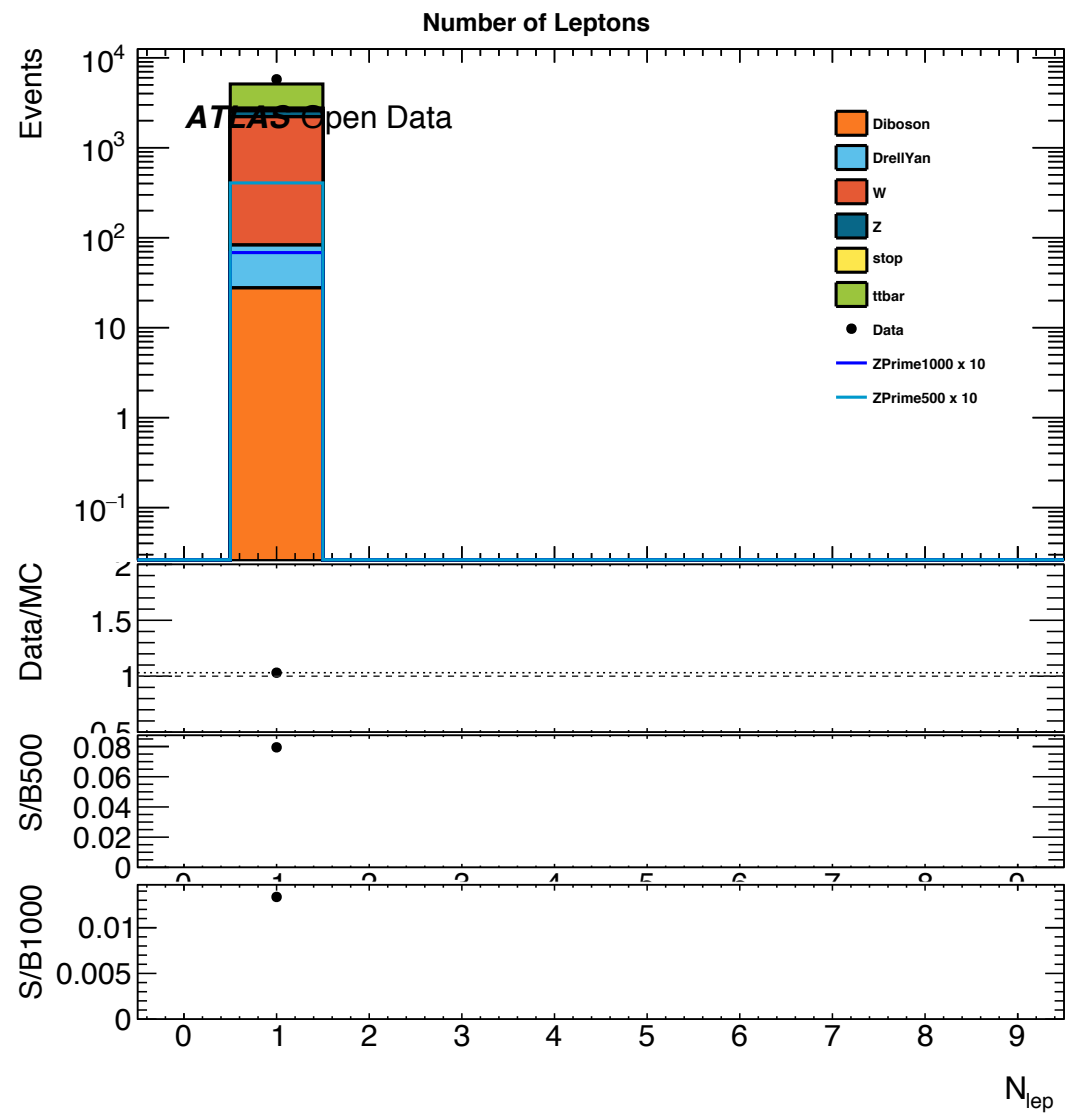
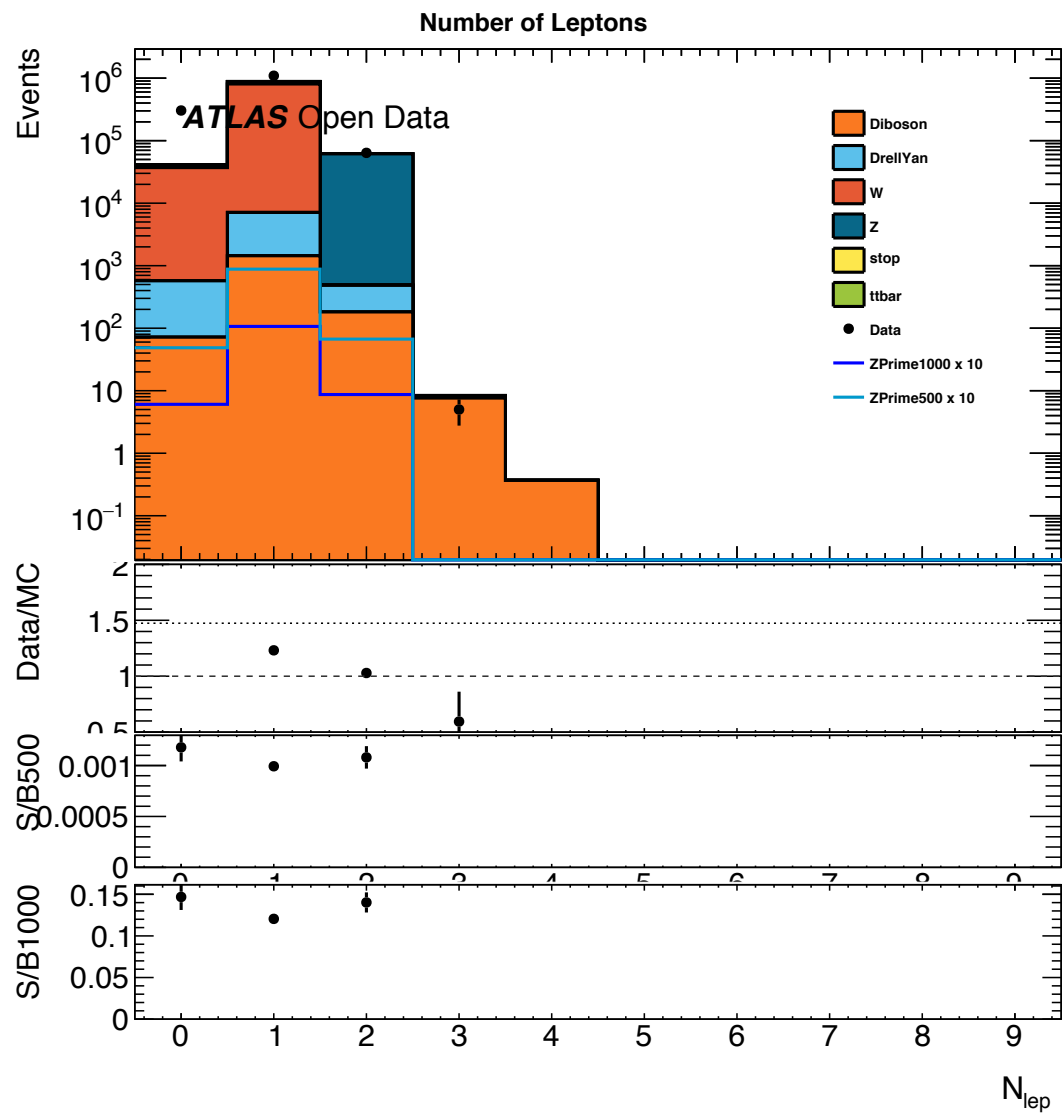
# Sumário

- Análise do decaimento do  $Z'$ , aplicando os cortes necessários
- Reconstrução do  $Z'$  a partir das partículas do decaimento
- Escolha das melhores features para utilizar em Machine Learning
- Exportação para um ficheiro csv
- Treino e validação da rede neuronal
- Estudo das correlações entre variáveis

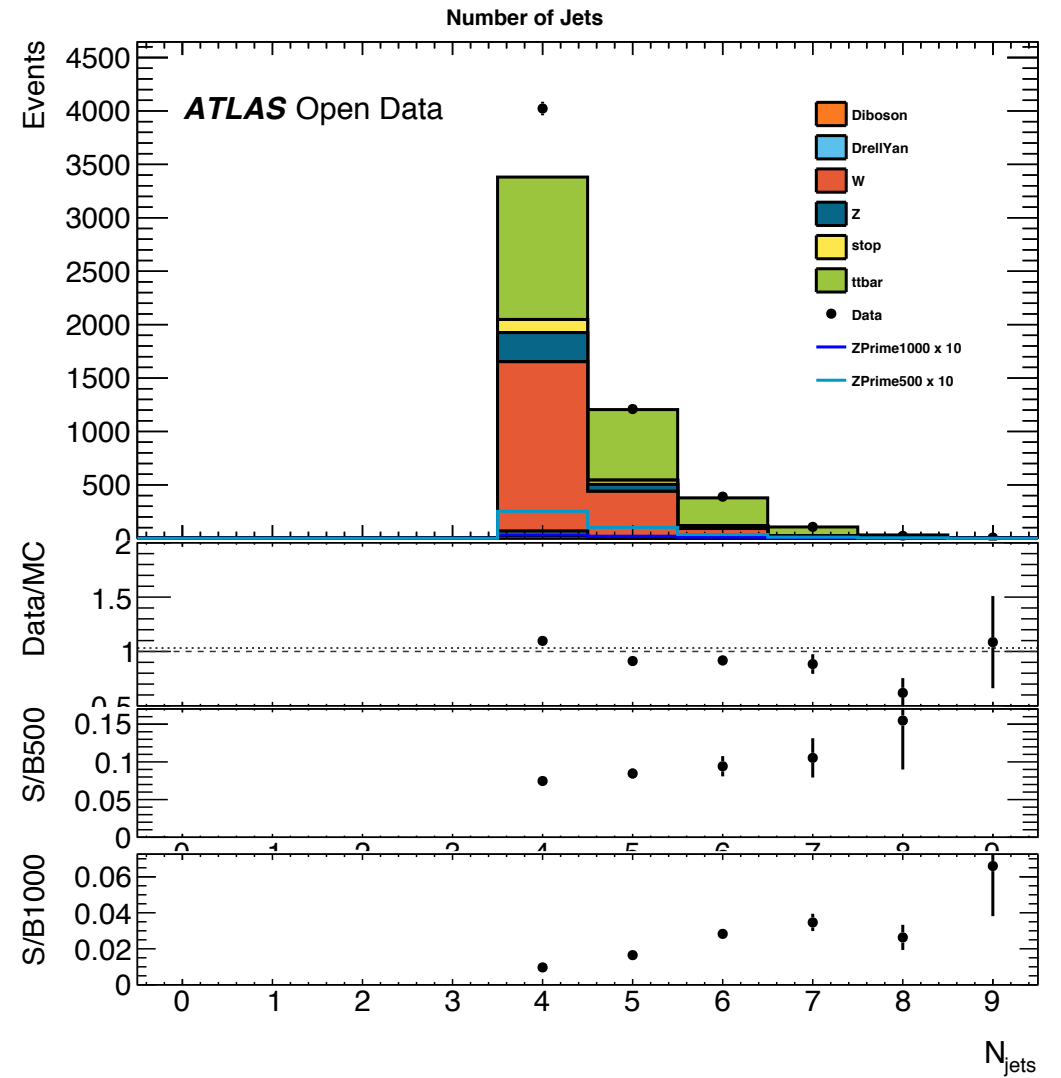
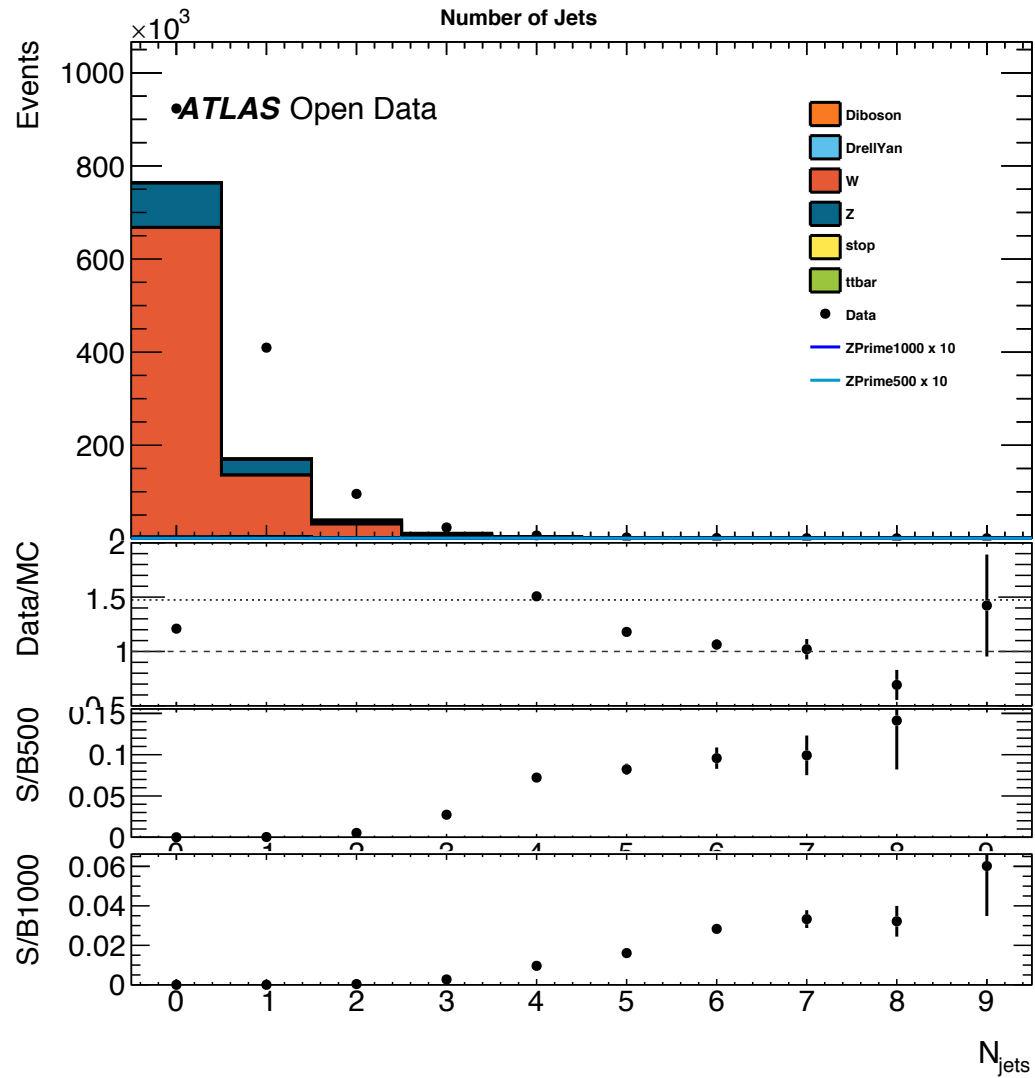
# Análise (cortes)

- Standard Cuts
- Corte no número de leptões = 1
- Corte no número mínimo de jatos  $\geq 4$
- Corte no número mínimo de quarks  $\geq 2$

# Leptões



# Jatos



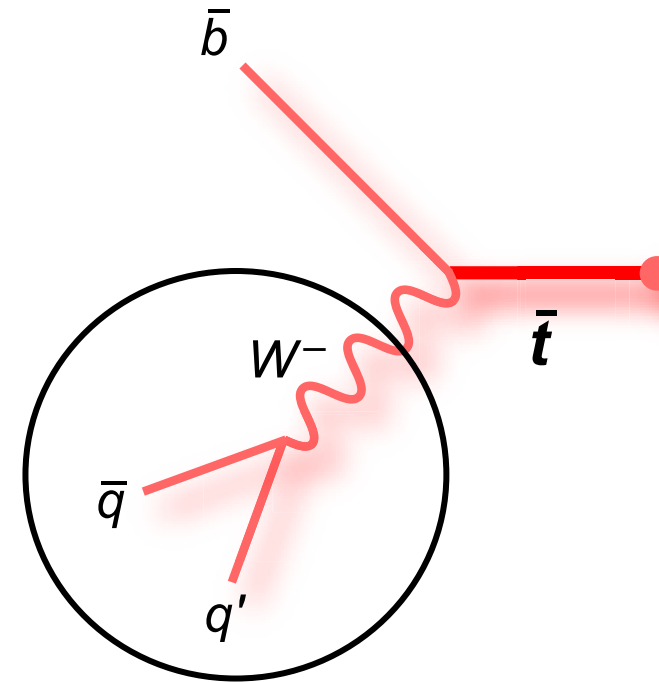
# Reconstrução do Z Prime

- Reconstrução do W hadrónico
  - Construir o TLorentzVector dos quarks
- Reconstrução do W leptónico
  - Construir o TLorentzVector do leptão e neutrino (problema: o momento em z do neutrino está em falta)
- Reconstrução do Top hadrónico e leptónico a partir dos W e b's
- Reconstrução do Z' a partir dos Tops

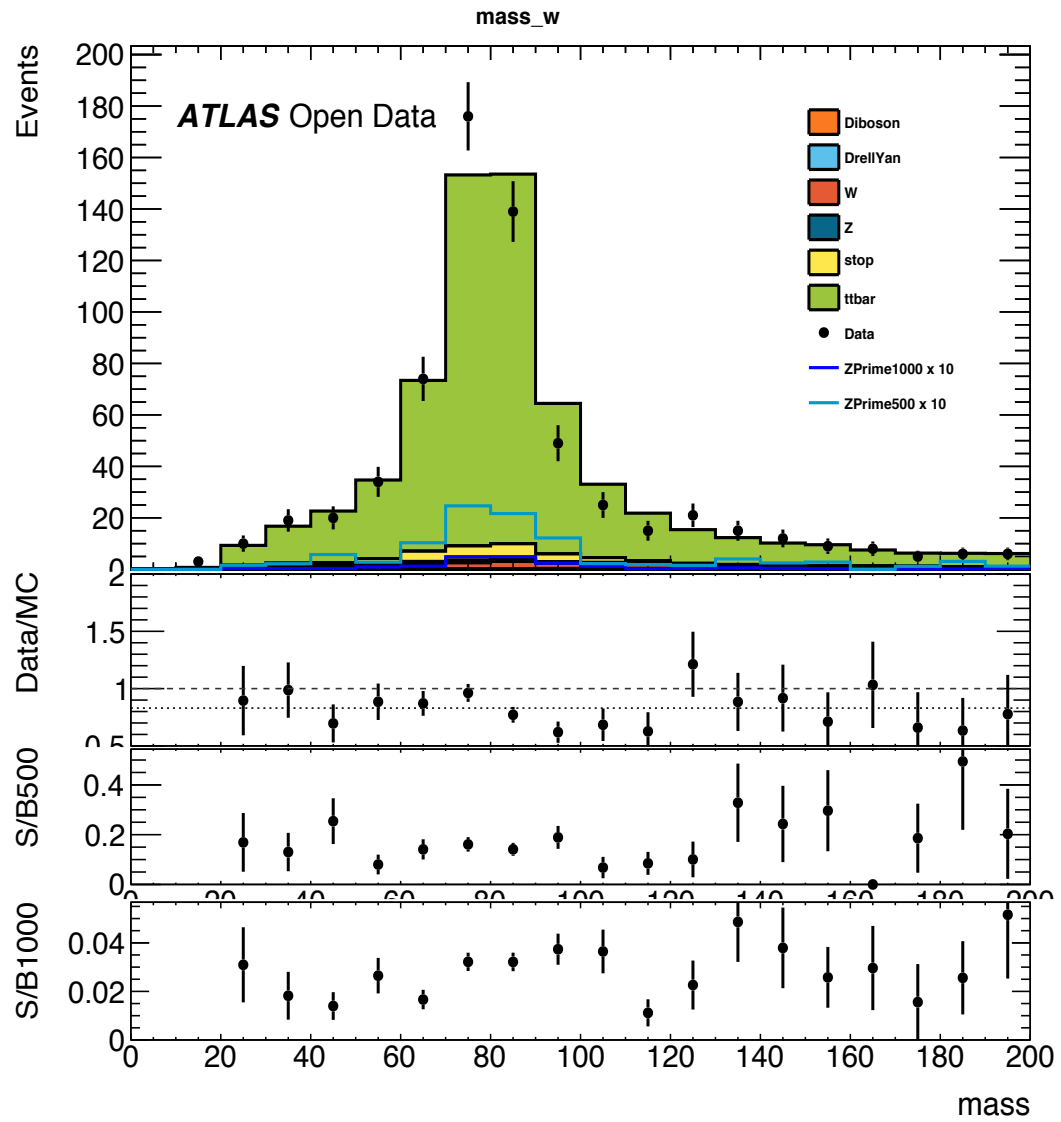
# Reconstrução W hadrónico

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- TLorentzVector dos quarks
- Combinatórias da soma de dois TLorentzVector dos quarks
- Diferença entre a massa do W e a massa das combinações
- O mínimo corresponde ao W reconstruído
- Elimina-se estes quarks da lista para evitar usar novamente noutra reconstrução

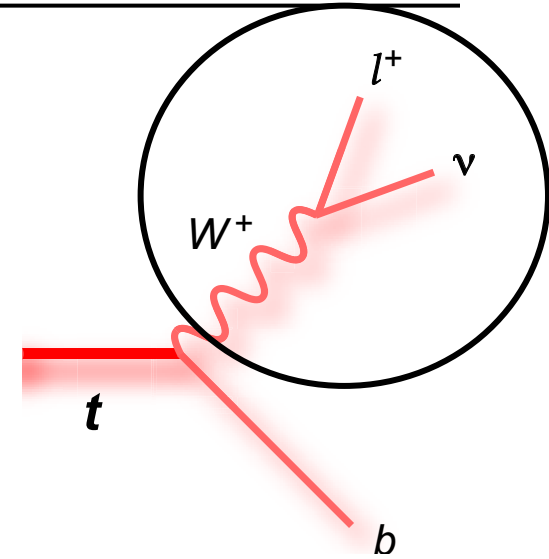






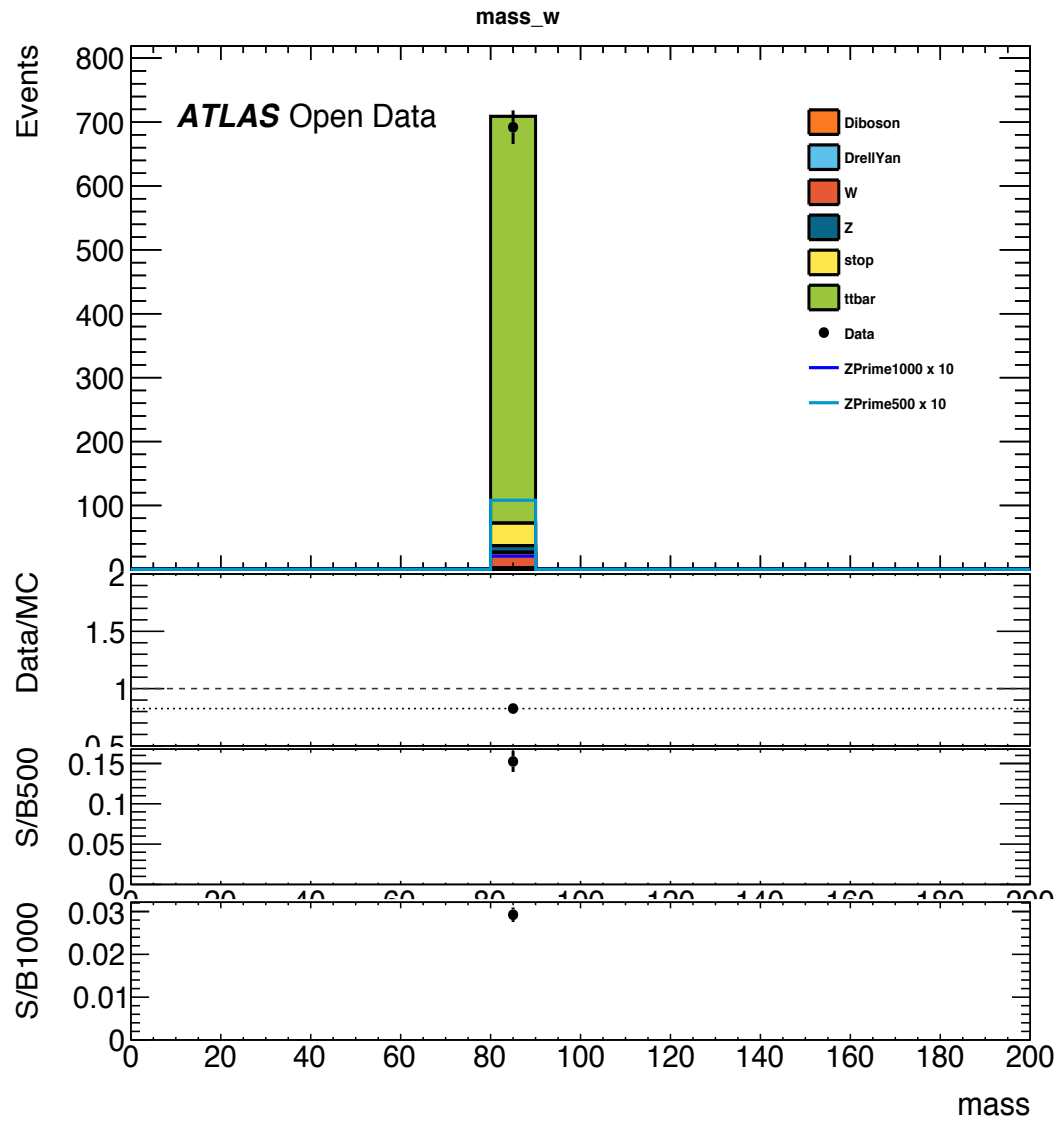
# Reconstrução W leptónico

- Para sabermos o momento em z do neutrino:



$$p_{\nu z} = \frac{2\vec{p}_T^{lep} \cdot \vec{p}_T^{\nu} p_z^{lep} + m_W^2 p_z^{lep}}{2(p_T^{lep})^2} \pm \frac{p^{lep} \sqrt{-4(p_T^{lep})^2 (p_T^{\nu})^2 + 4(\vec{p}_T^{lep} \cdot \vec{p}_T^{\nu})^2 + 4\vec{p}_T^{lep} \cdot \vec{p}_T^{\nu} m_W^2 + m_W^4}}{2(p_T^{lep})^2}$$

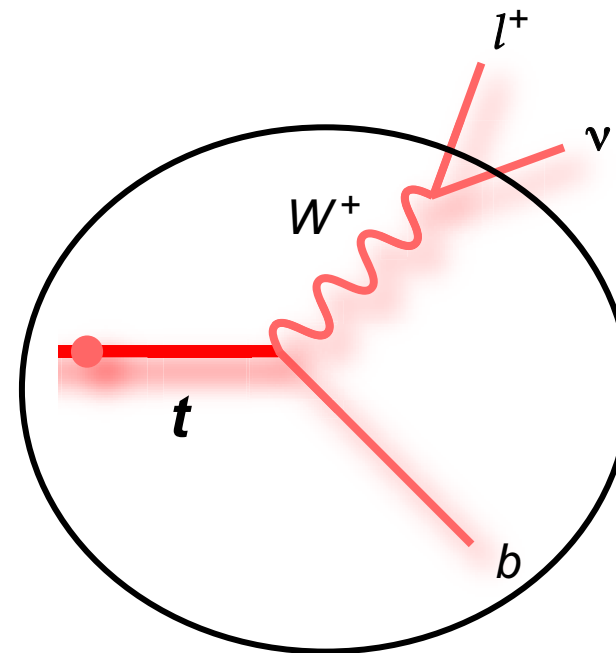
- TLorentzVector do leptão e do neutrino (a partir do momento em  $z$  e da energia em falta)
- Combinatórias da soma de dois TLorentzVector
- Diferença entre a massa do  $W$  e a massa das combinações
- O mínimo corresponde ao  $W$  reconstruído



# Reconstrução Top Leptônico

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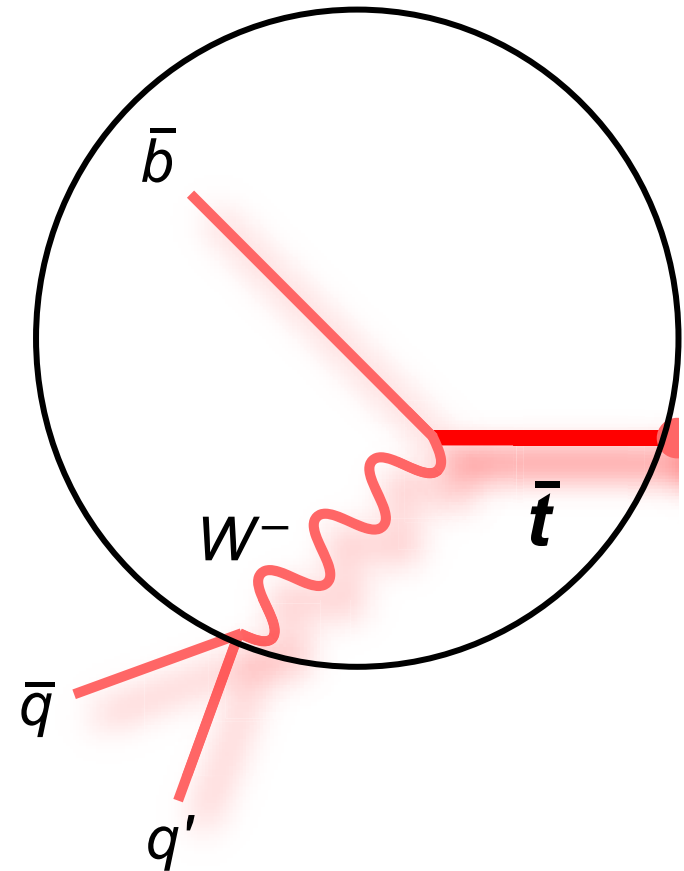
- Se nº btags maior que 2
  - TLorentzVector dos b's e do W
  - Soma dos TLorentzVector e usar o b que minimize a diferença entre a combinação e a massa do top
- Se nº btags menor que 2
  - TLorentzVector dos quarks e do W
  - Soma dos TLorentzVector e usar o quark que minimize a diferença entre a combinação e a massa do top



# Reconstrução Top Hadrónico

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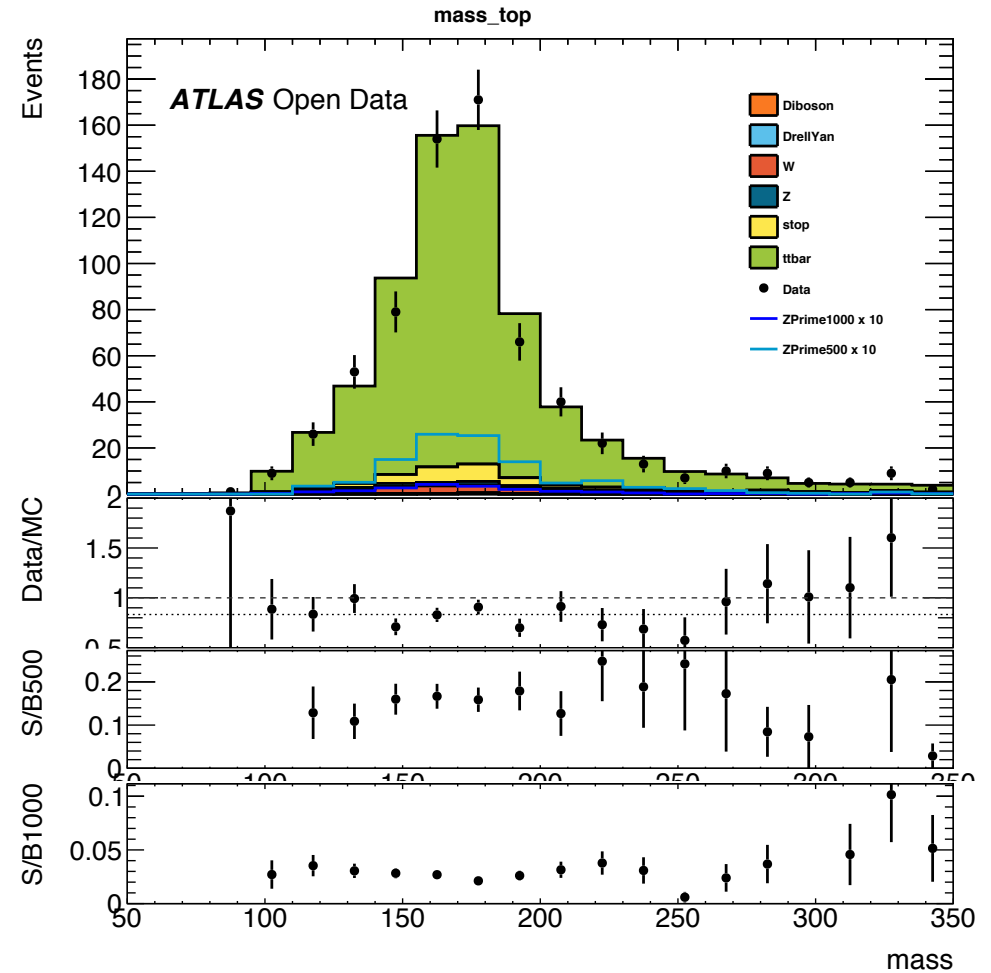
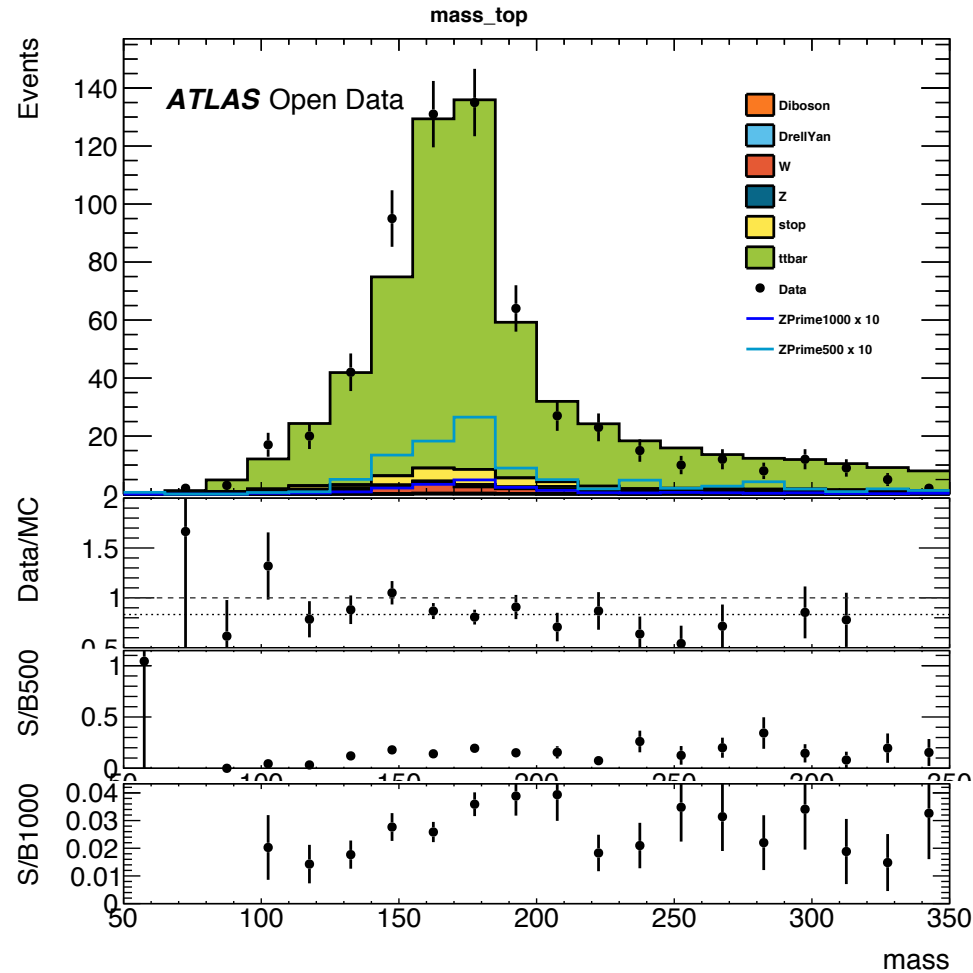
- Se nº btags maior que 2
  - TLorentzVector dos b's e do W
  - Soma dos TLorentzVector e usar o b que minimize a diferença entre a combinação e a massa do top
- Se nº btags menor que 2
  - TLorentzVector dos quarks e do W
  - Soma dos TLorentzVector e usar o quark que minimize a diferença entre a combinação e a massa do top



- Se forem utilizados os mesmos b's ou quarks para ambos os tops
  - O segundo b (ou quark) que minimize a diferença entre a massa do top e as combinações deverá ser utilizado

# Massa Top Had

# Massa Top Lep

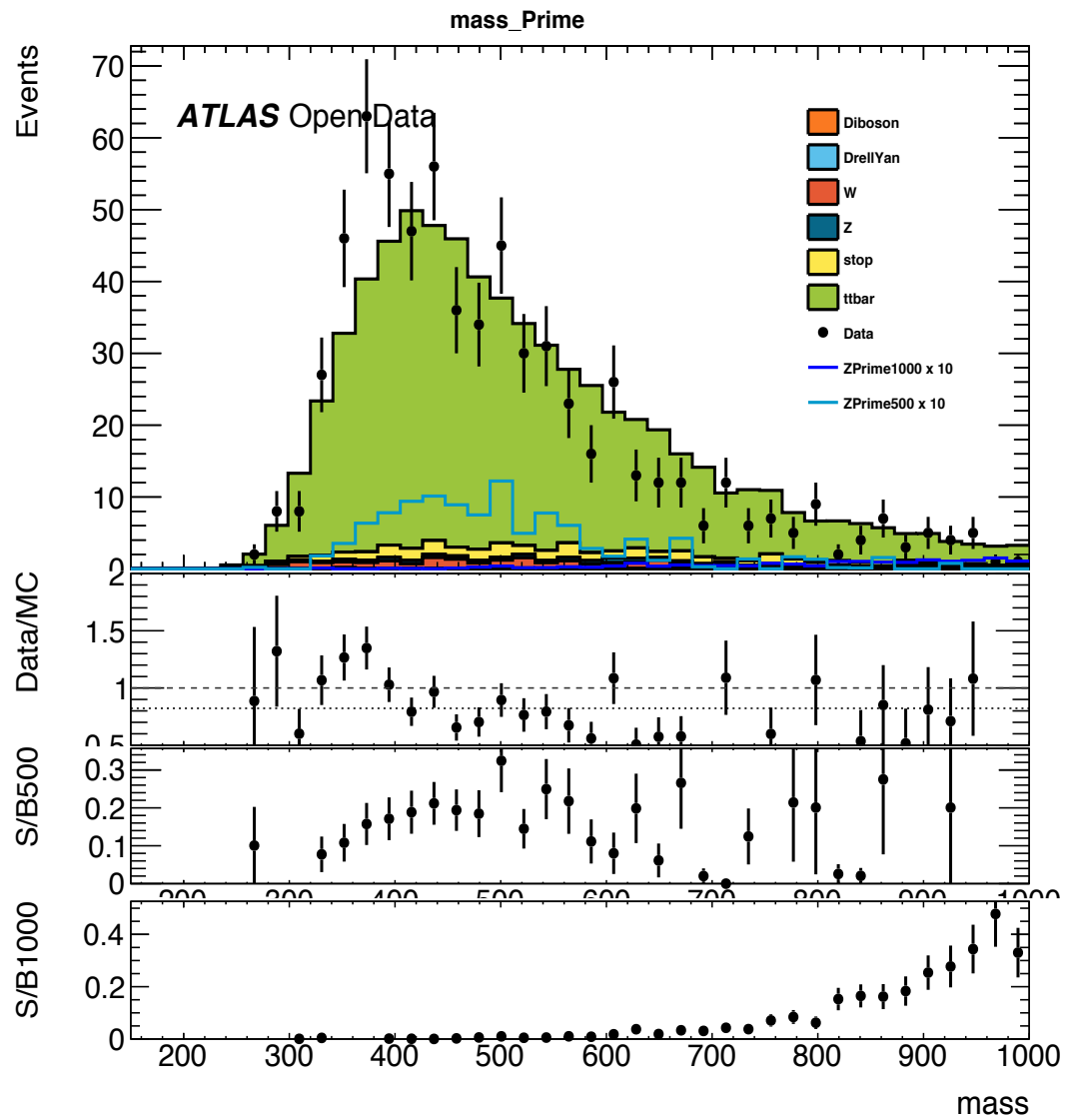




# Reconstrução do Z Prime

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- Somou-se os TLorentzVector dos dois tops



# Features

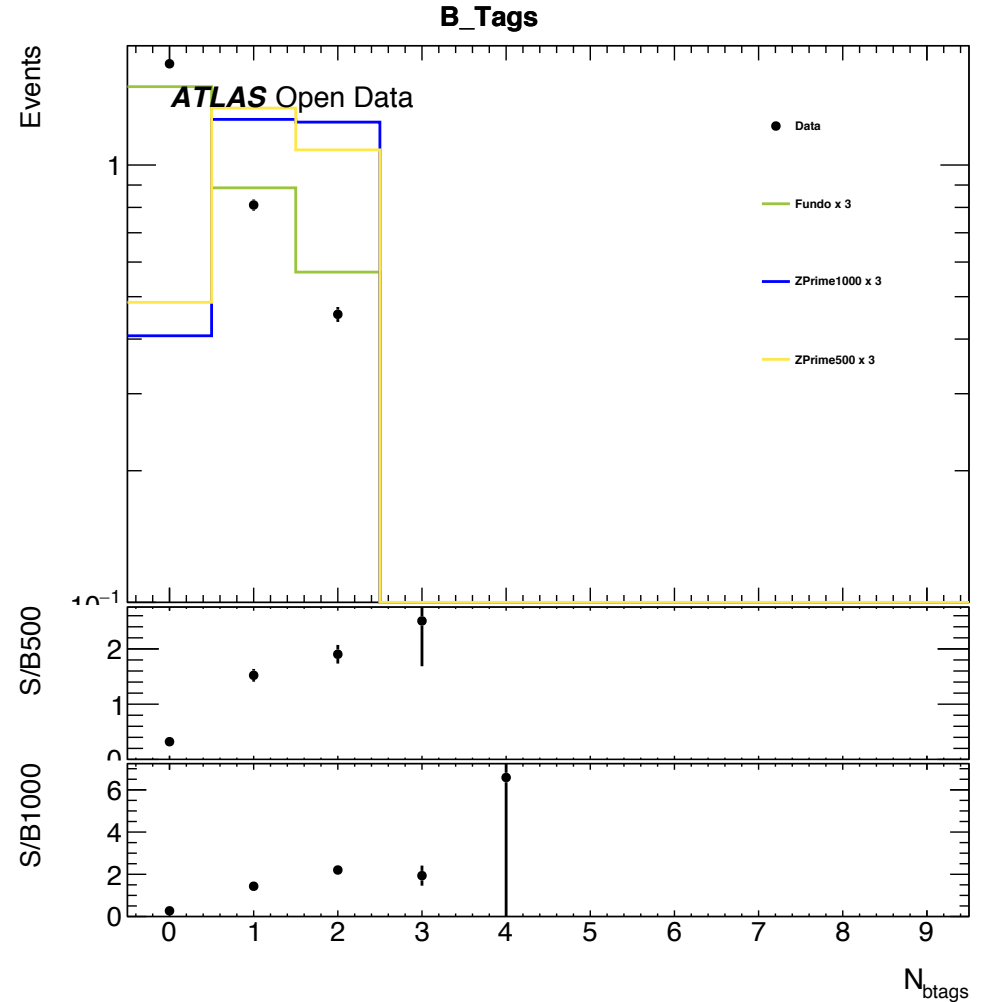
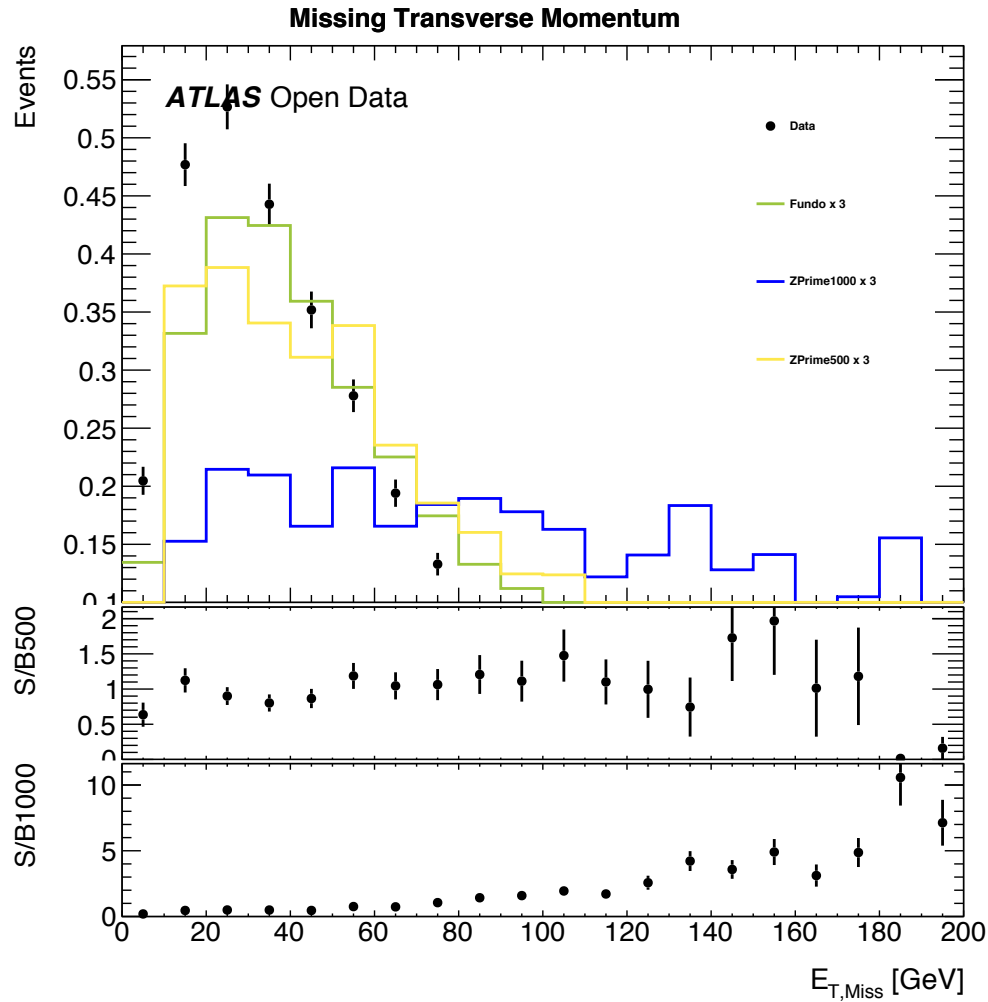
- Baixo nível:
  - Energia em falta (E<sub>miss</sub>)
  - Número de quarks b
  - Pseudorapidez do leptão
  - Pseudorapidez dos 4 jatos mais energéticos
  - Delta R
  - Delta Phi (entre jatos)
  - Delta eta (entre jatos)
- Alto nível:
  - Massa do Z Prime
  - Momento transversal do Z Prime
  - Soma escalar dos momentos transversos dos leptões e jatos finais (incluindo quarks b)

$$\Delta\eta = \sqrt{\Delta\Phi^2 + \Delta\eta^2}$$

$$\Delta\Phi = \phi_1 - \phi_2$$

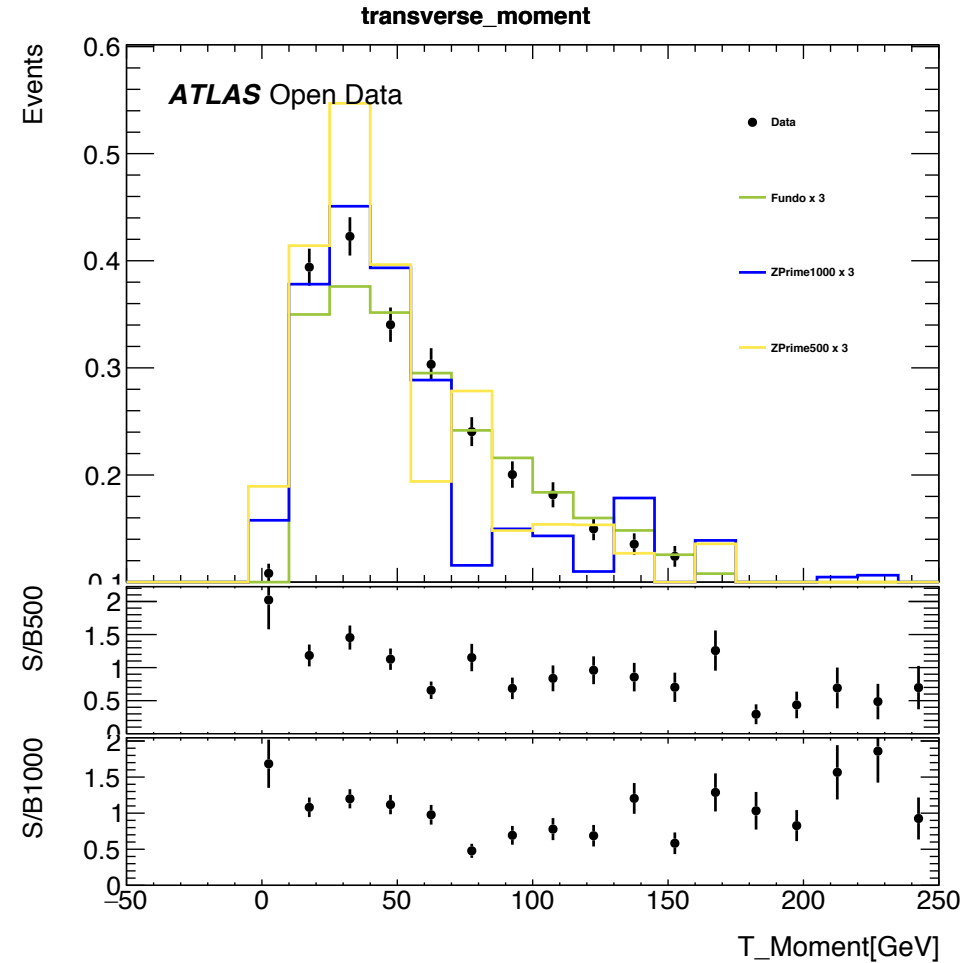
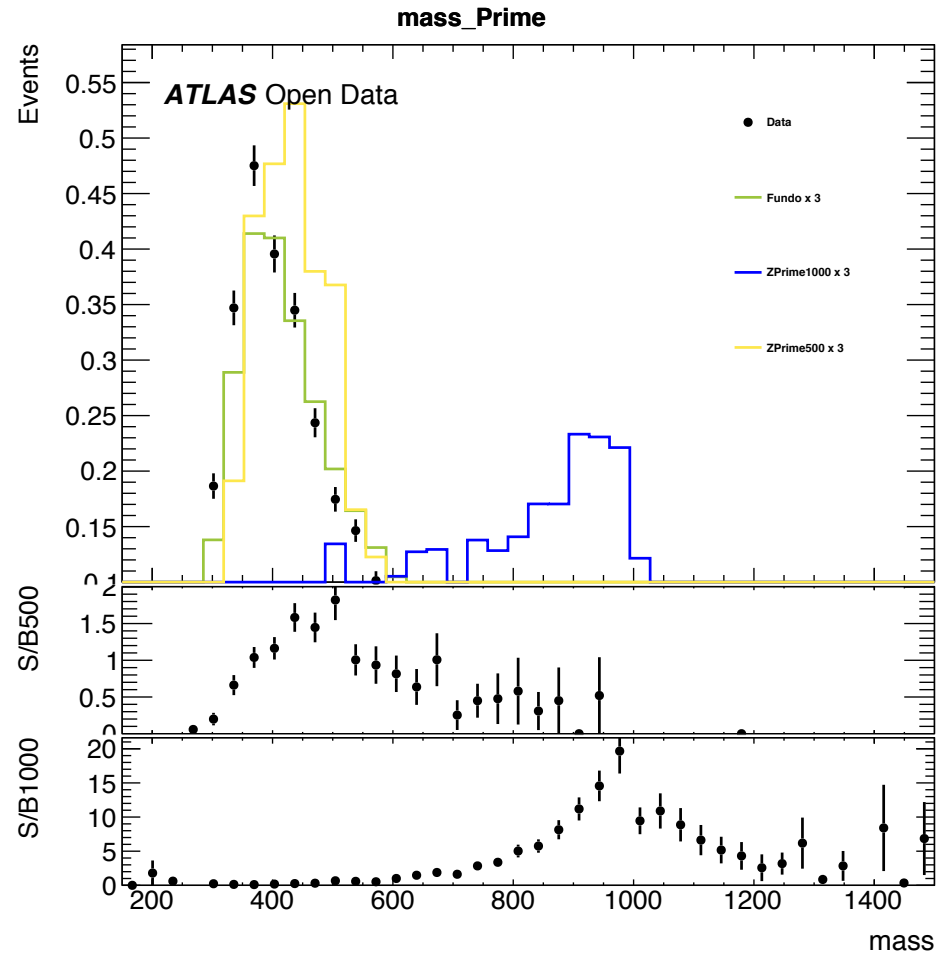
# Etmiss

# BTags



# Mass ZPrime

# PT ZPrime



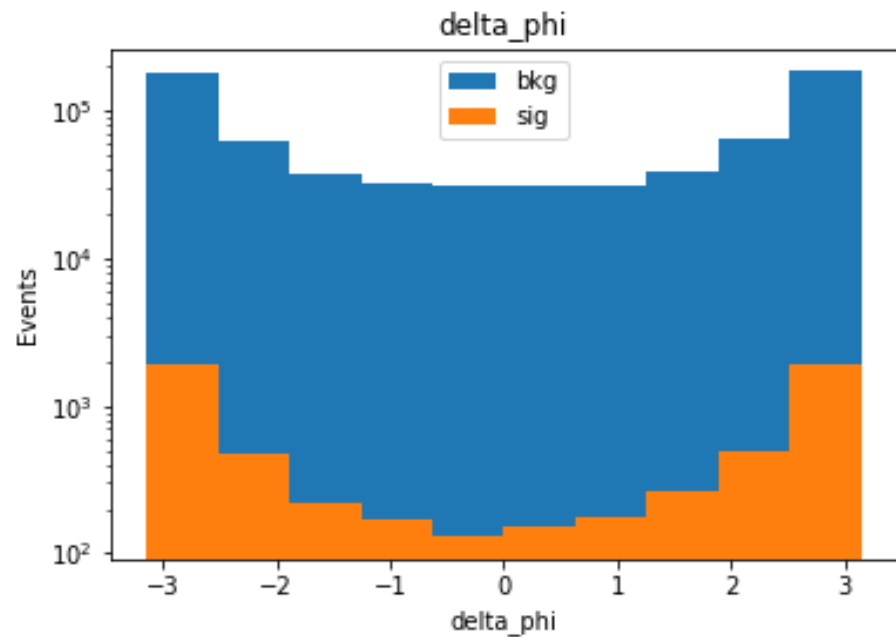
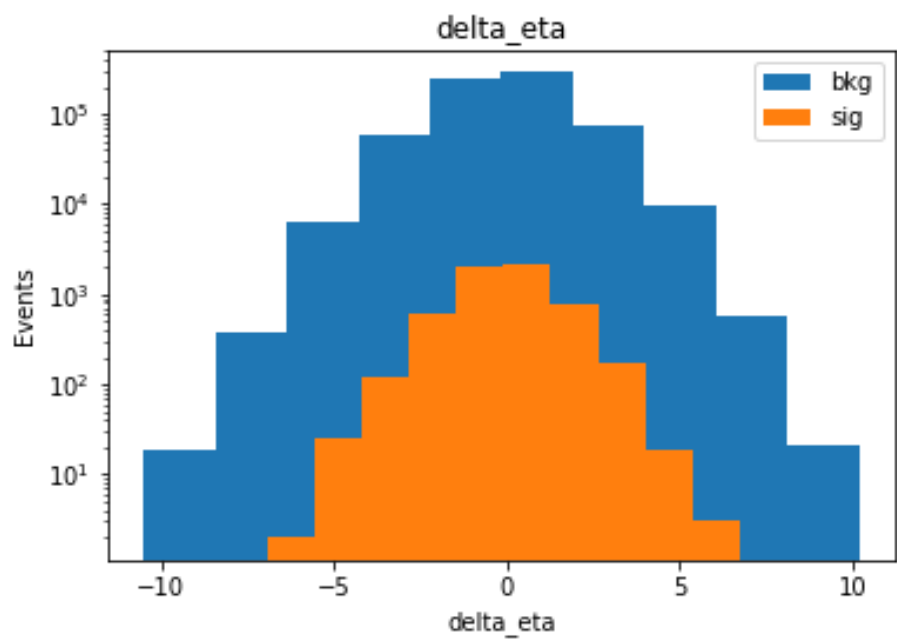
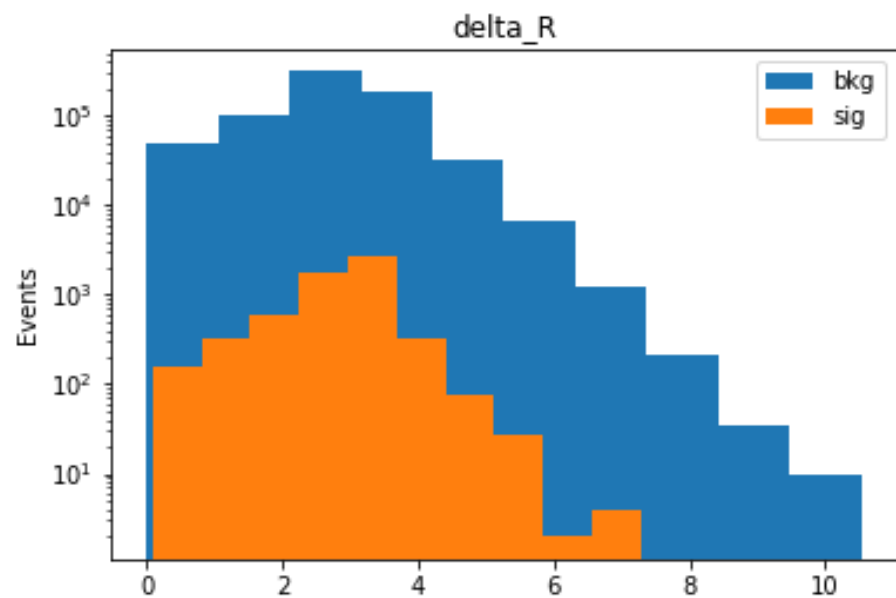
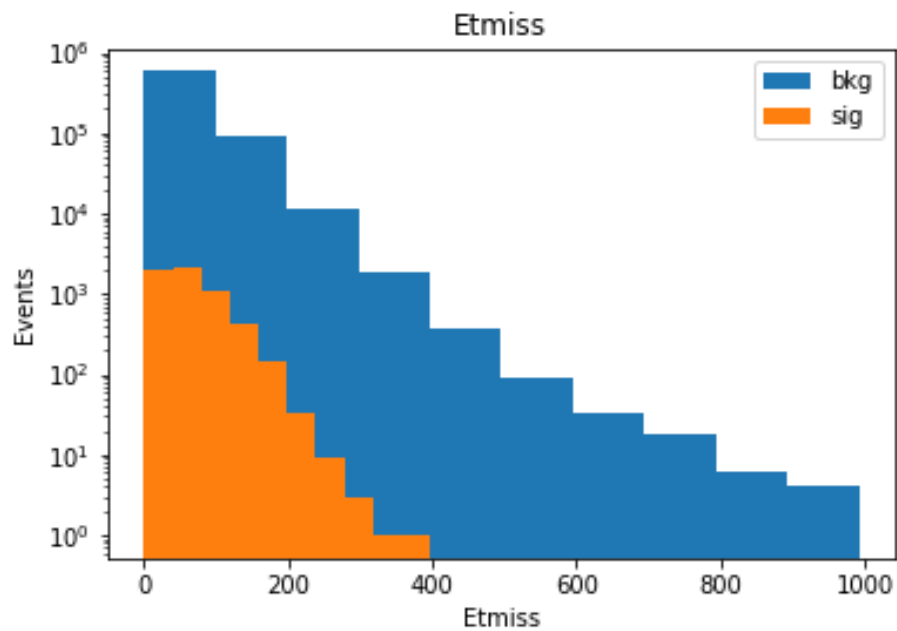
# Treino e validação da rede neuronal

```
model = Sequential()  
model.add(Dense(100, activation='relu', input_shape=X_train.shape[1:]))  
model.add(Dense(100, activation='relu'))  
model.add(Dense(1, activation='sigmoid'))
```

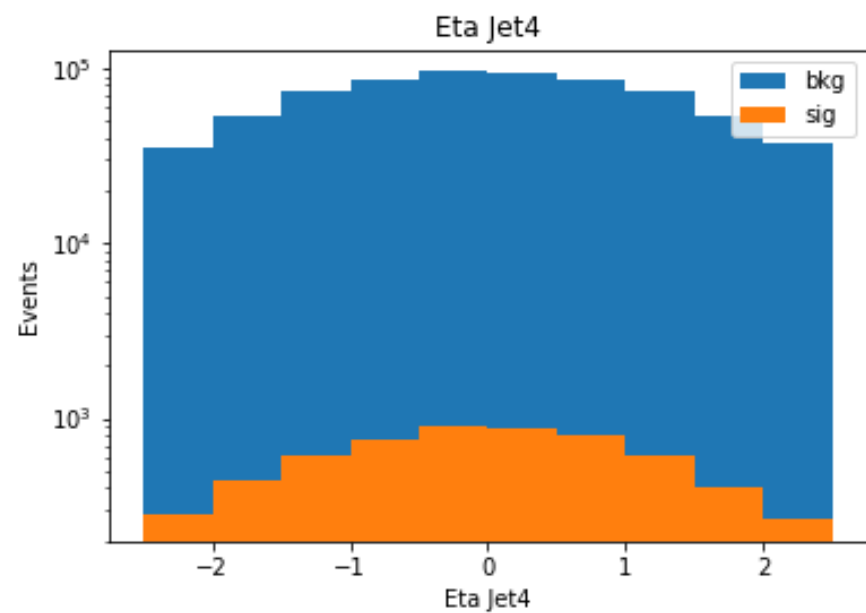
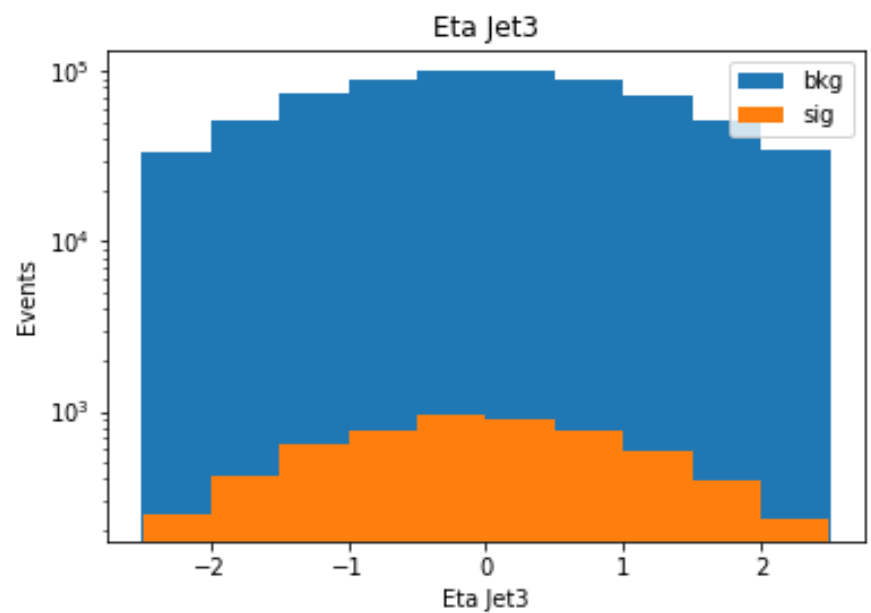
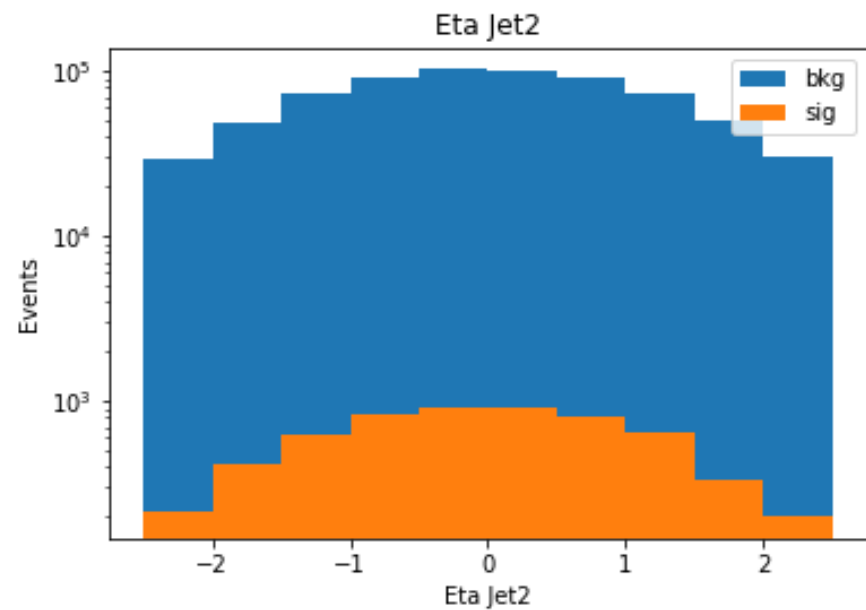
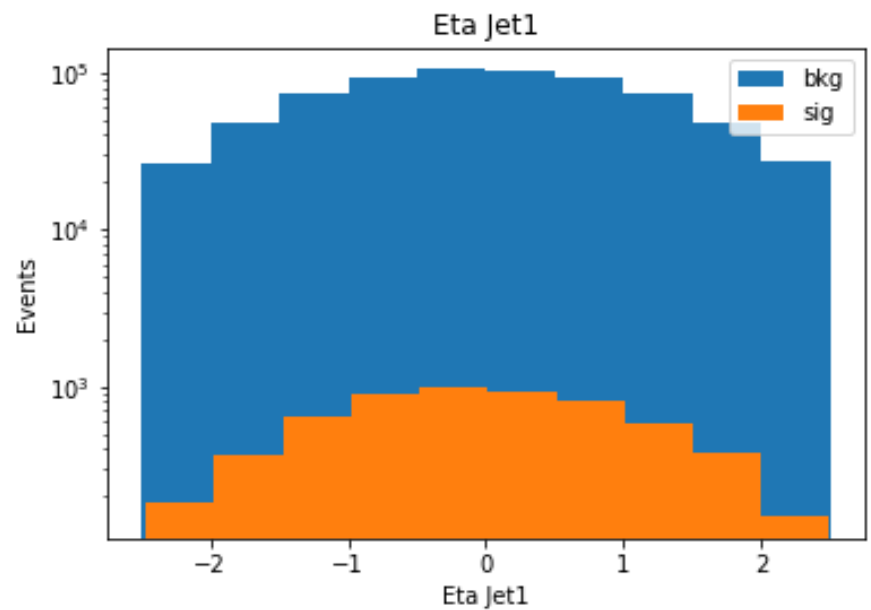
```
earlyStop = EarlyStopping(monitor='val_loss', patience=5, verbose=1, mode='auto') #Setup early stopping  
saveBest = ModelCheckpoint("train_weights/best.h5", monitor='val_loss', verbose=1, #Save best performing network  
                           save_best_only=True, save_weights_only=True, mode='auto', period=1)
```

```
adam = Adam(lr=0.00001, beta_1=0.9, beta_2=0.999, epsilon=None, decay=0.0, amsgrad=False)  
model.compile(loss='binary_crossentropy',  
             optimizer=adam)
```

# ZPRIME500

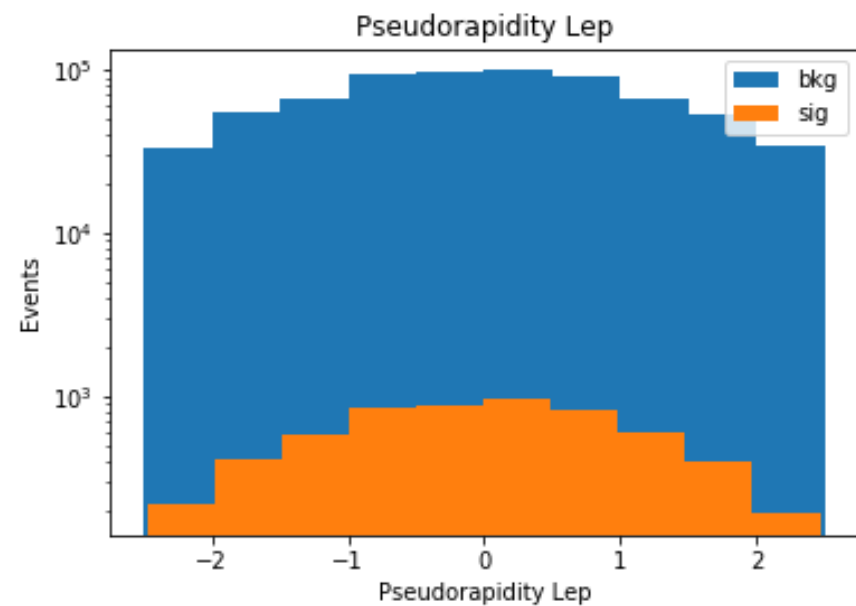
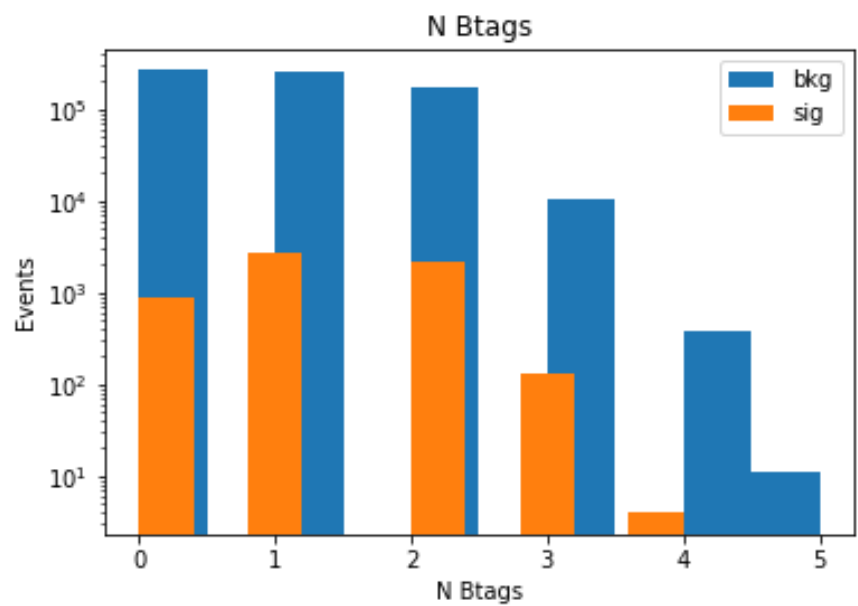
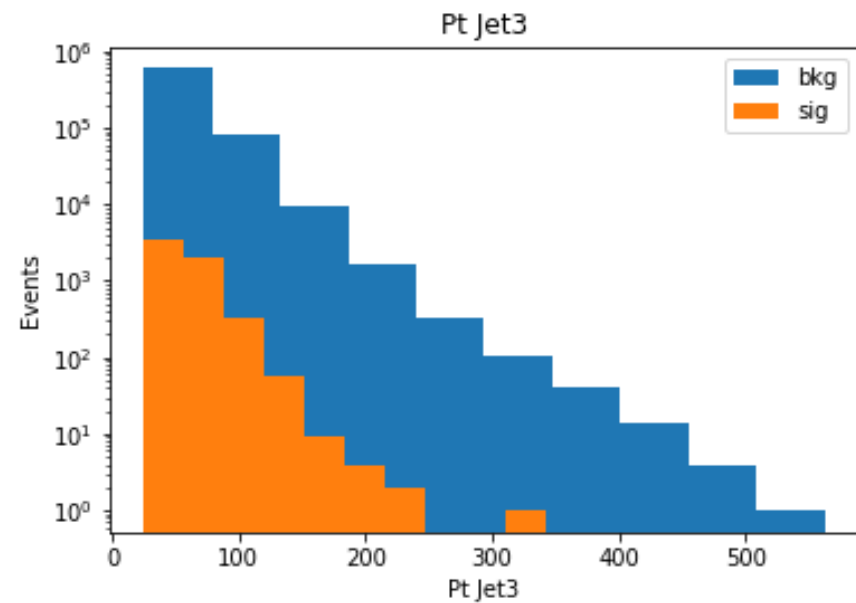
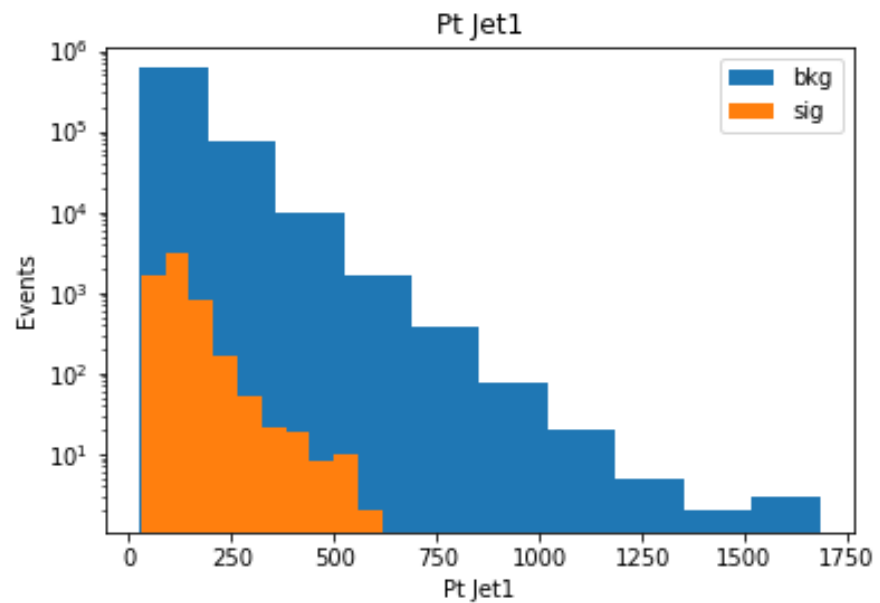


# ZPRIME500

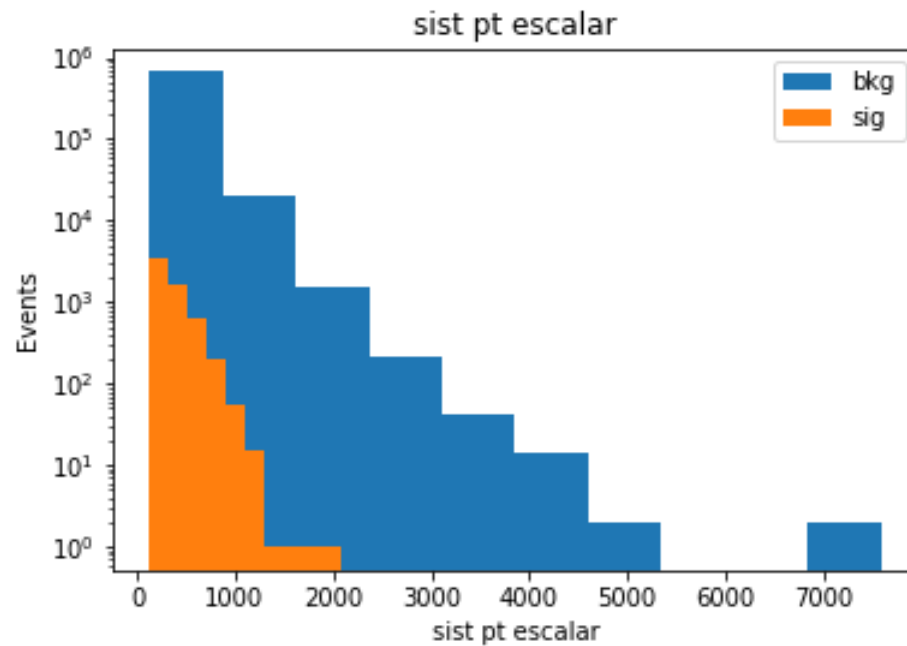
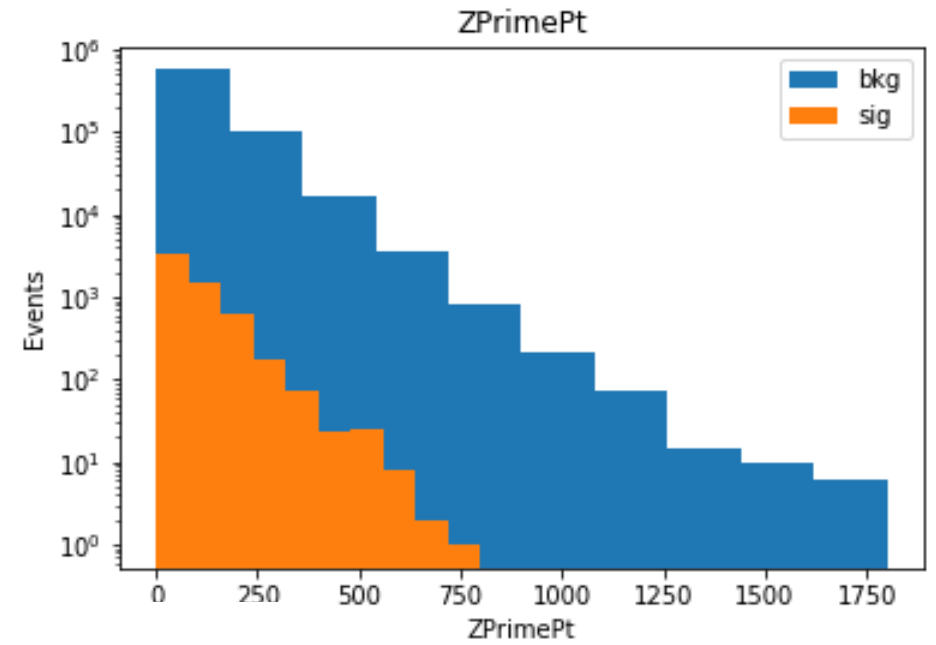
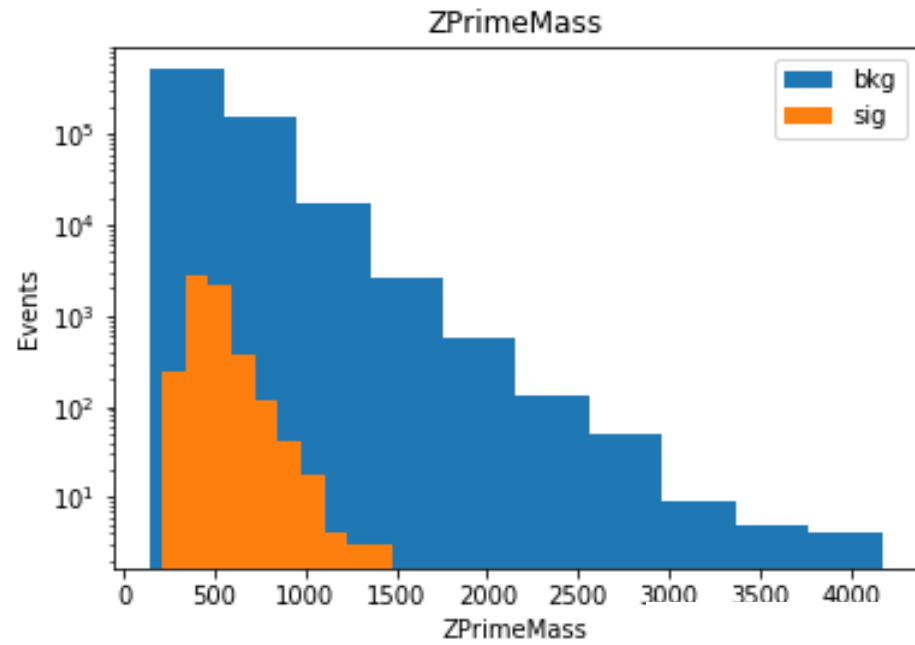




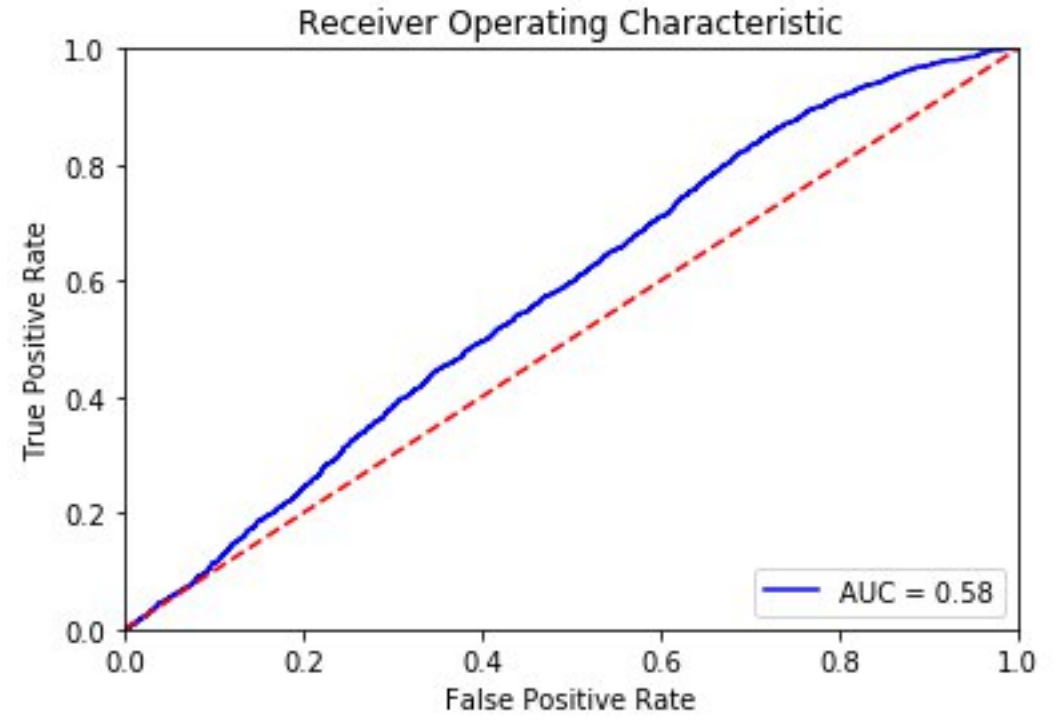
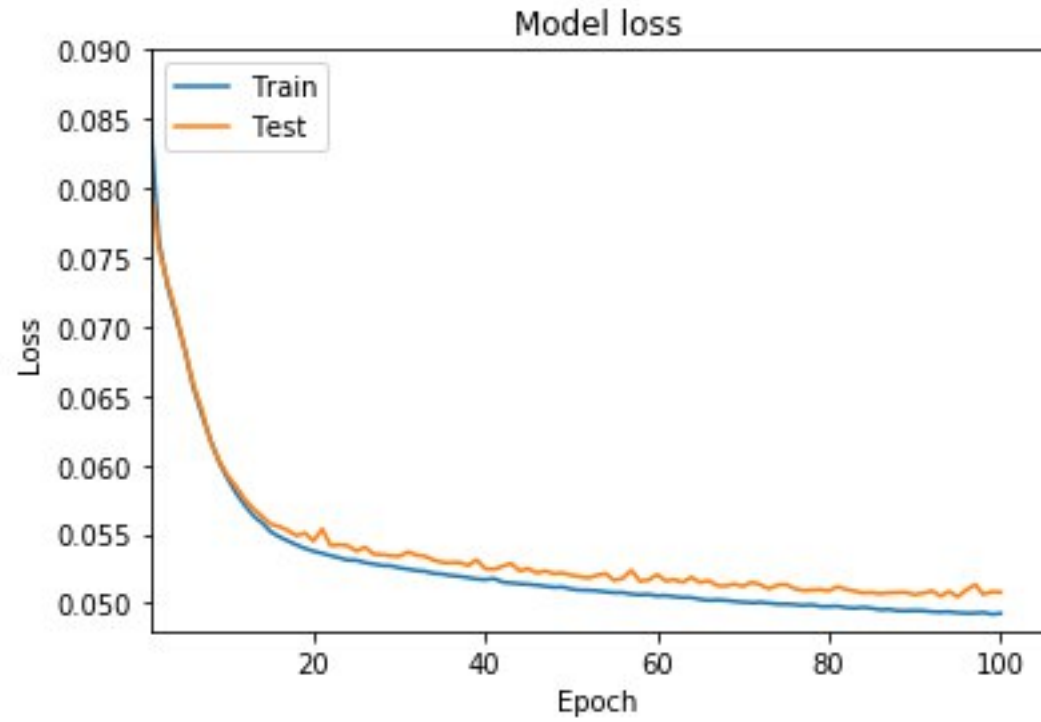
# ZPRIME500



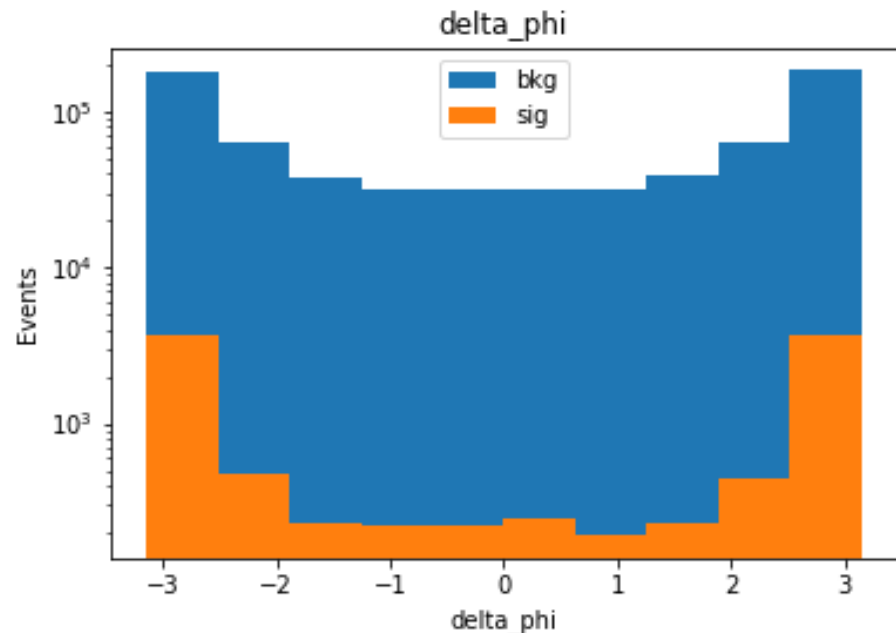
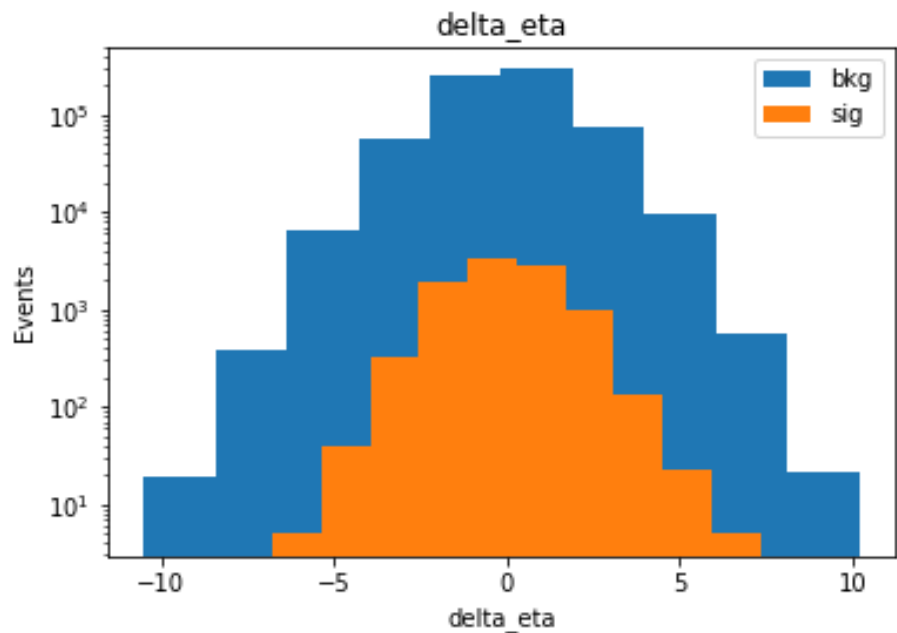
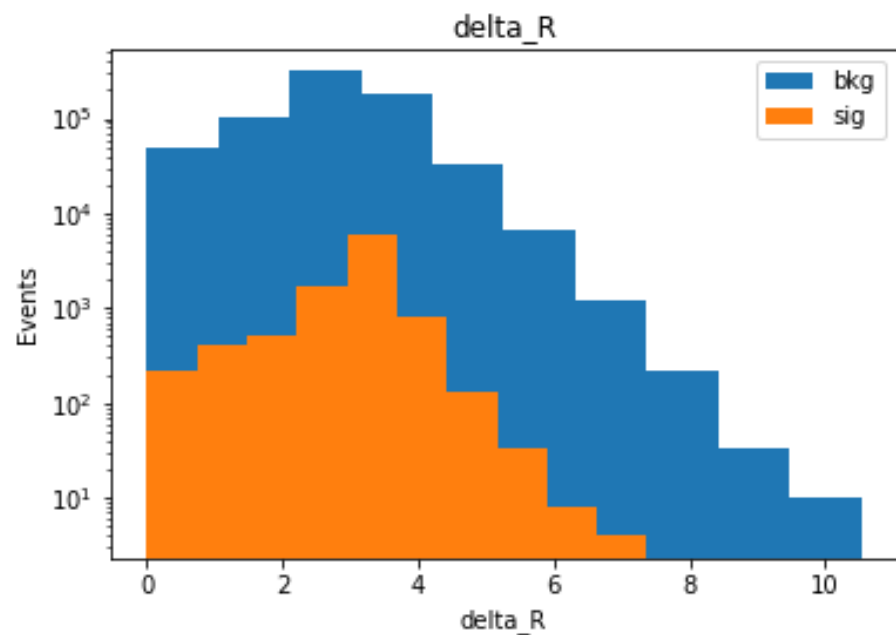
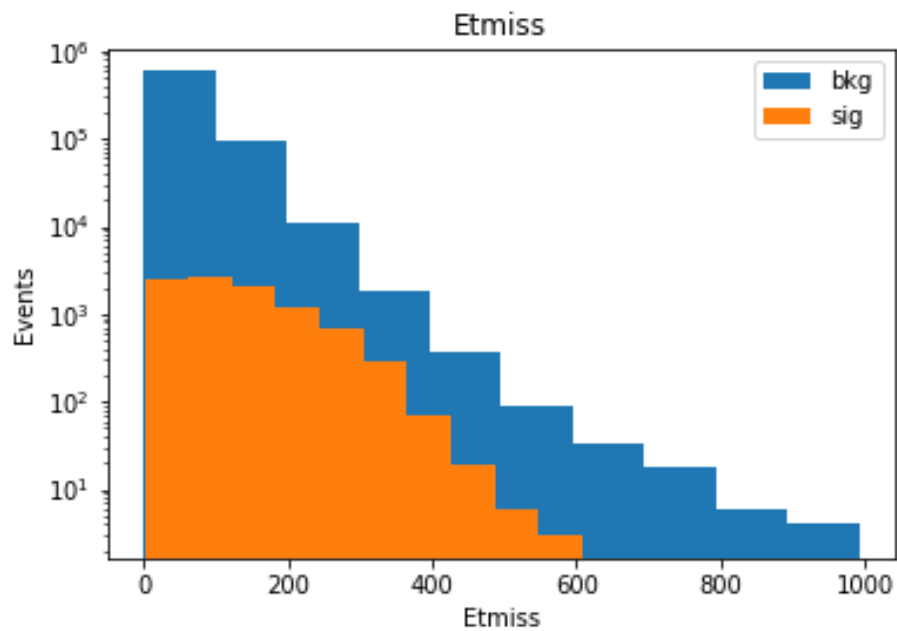
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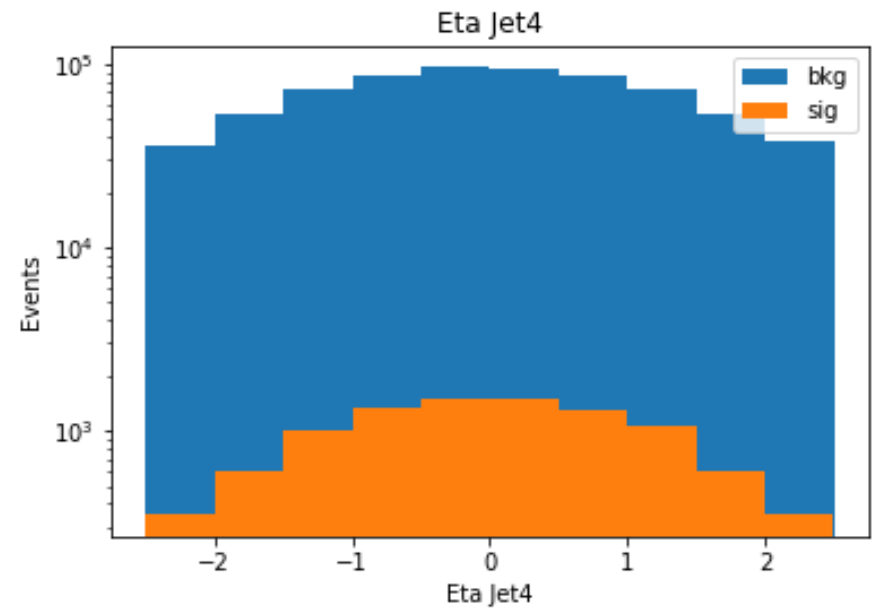
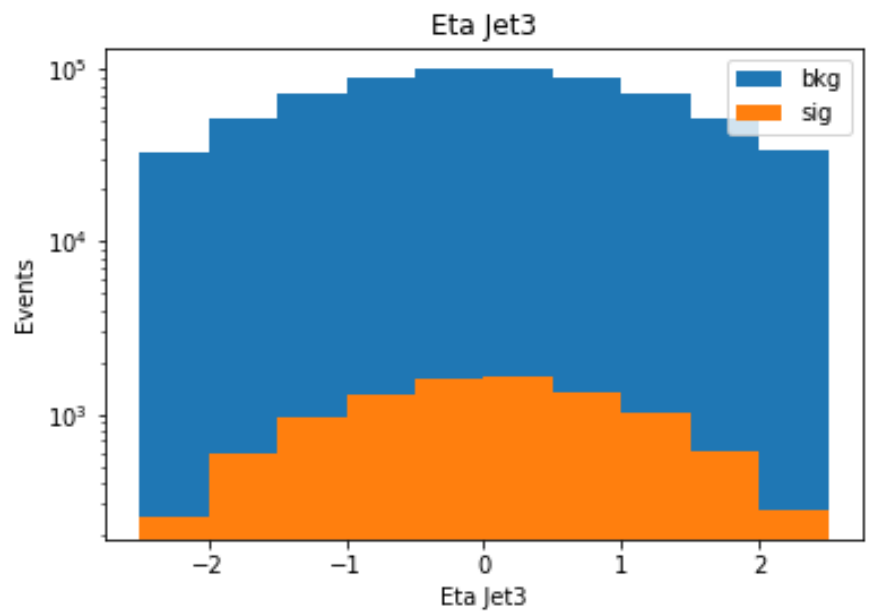
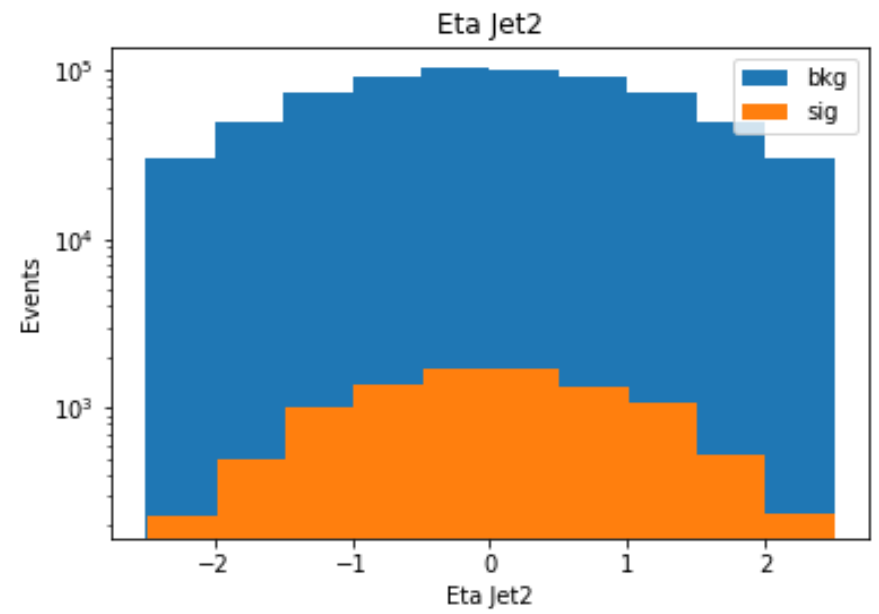
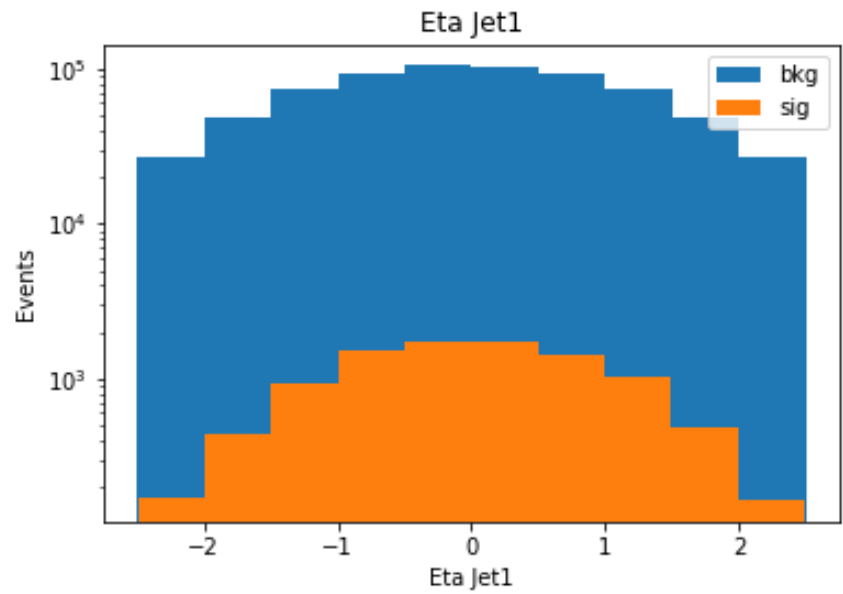
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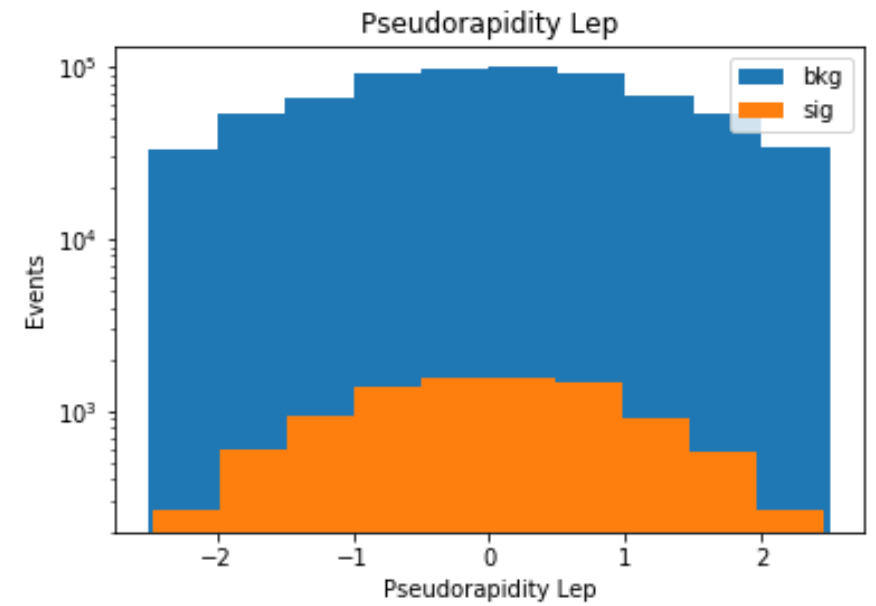
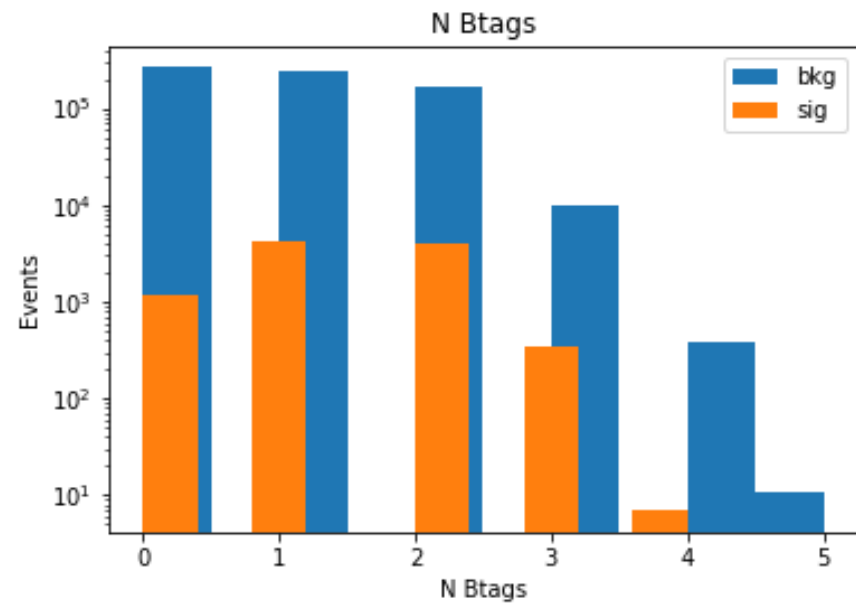
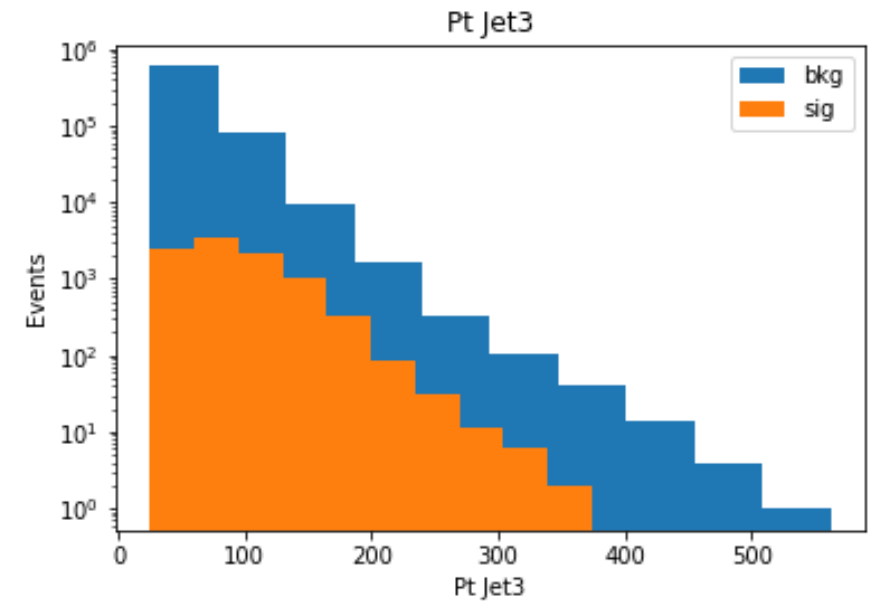
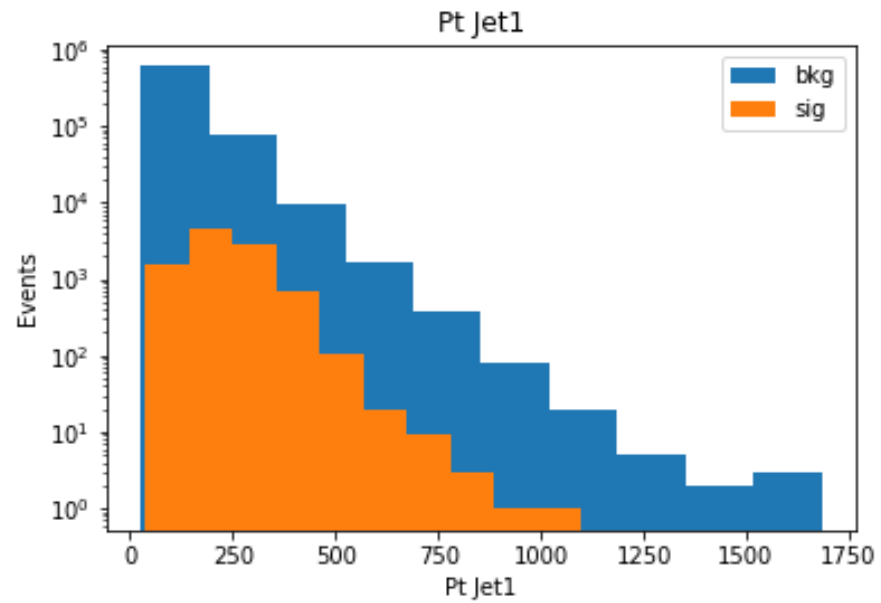
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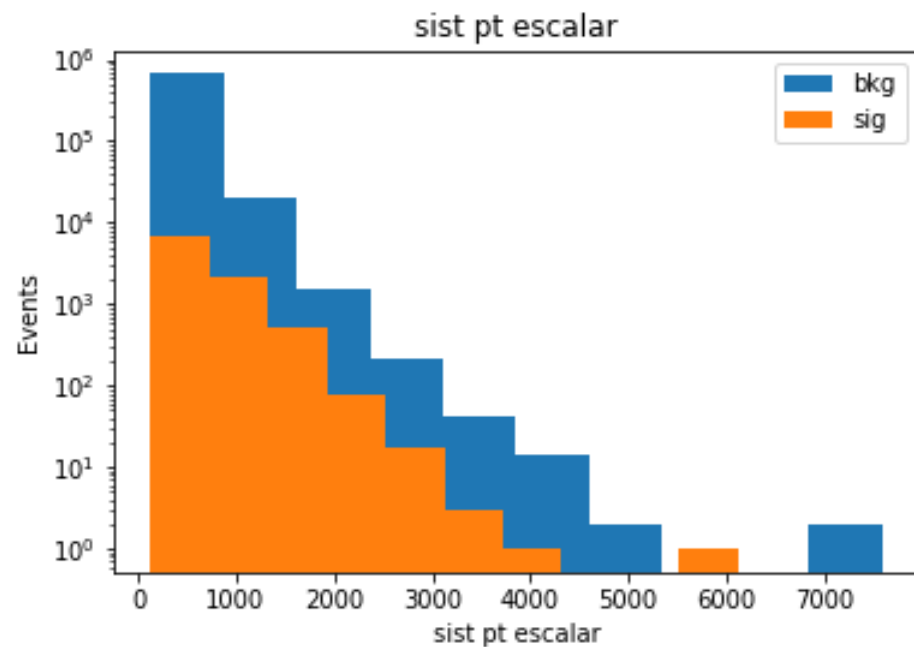
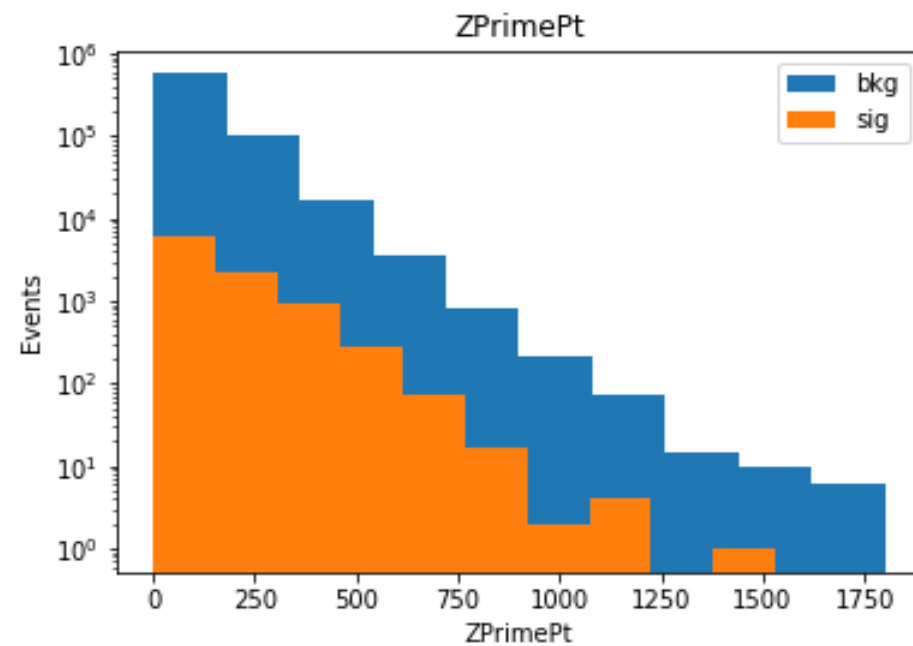
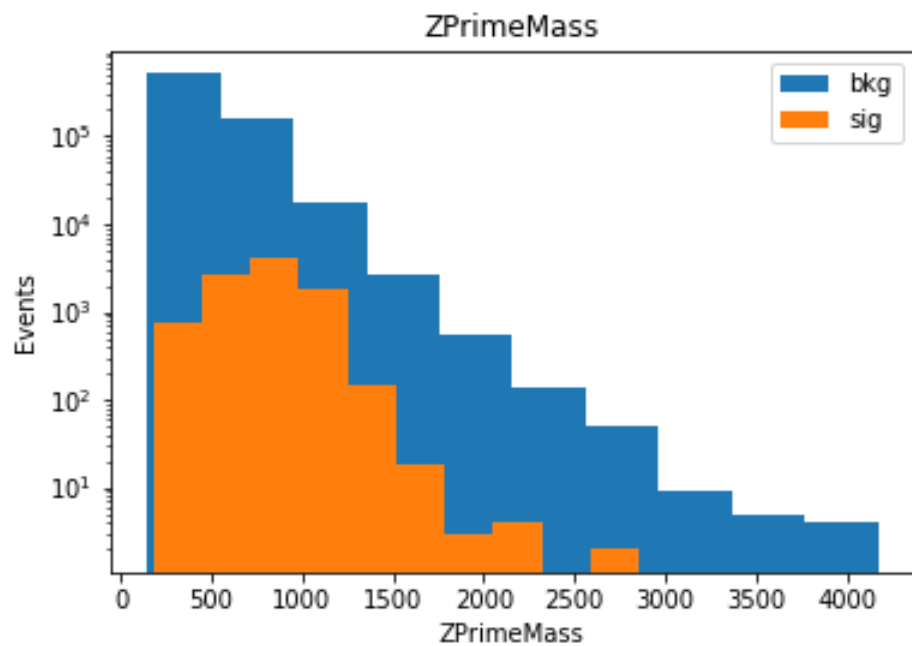
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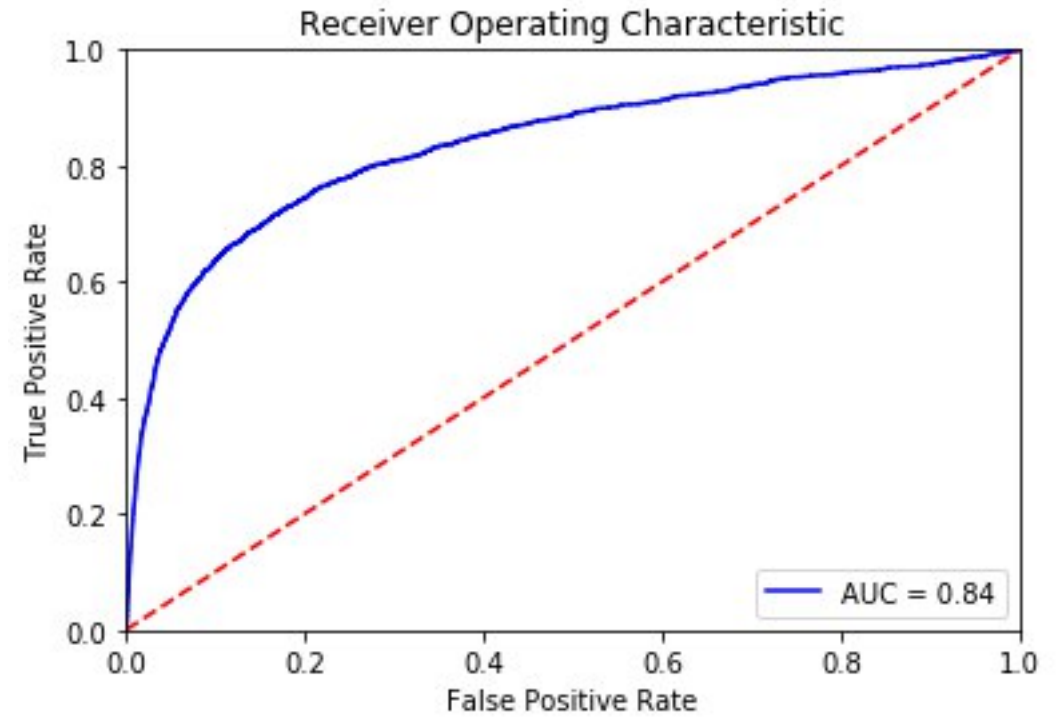
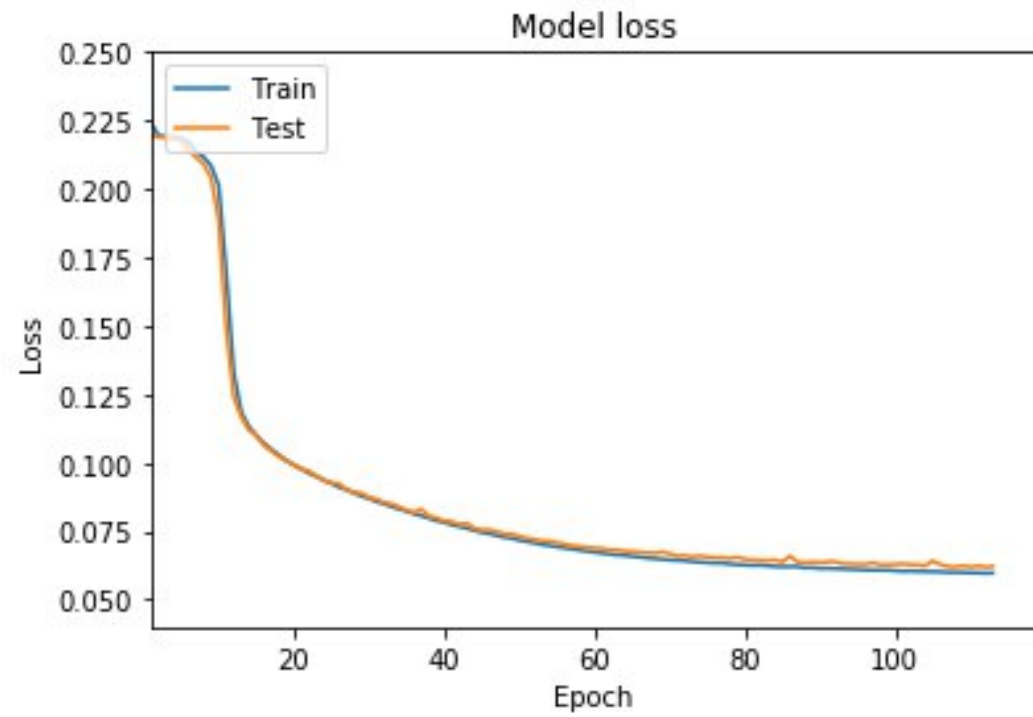
# ZPRIME1000



# ZPRIME1000



# ZPRIME1000











## Próximos objetivos

- Calcular limites a partir do output do classificador