

# Results and future outlook for the $H \rightarrow Z\gamma$ analysis on the ATLAS experiment.

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November 2025



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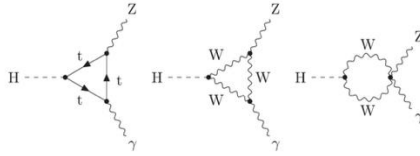




# Overview

Analysis Objective:  $H \rightarrow Z\gamma \rightarrow \ell\ell\gamma$

- a potential channel to explore physics **Beyond the Standard Mode**
- rare decay only via loop diagrams in the SM



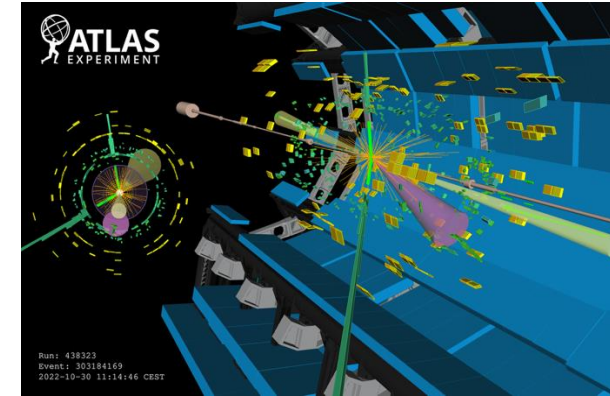
Partial Run 3 Data: **13.6 TeV** and integrated luminosity of **165 fb<sup>-1</sup>**

Irreducible background

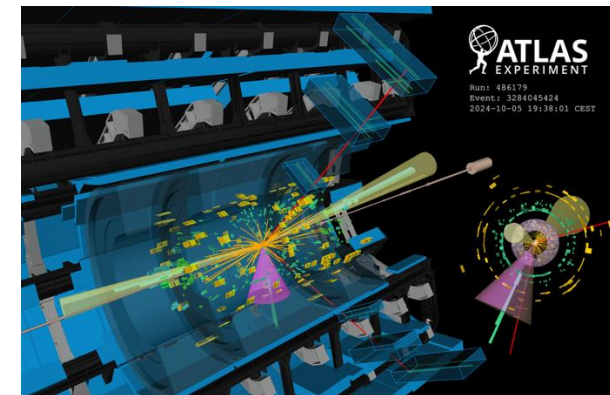
- the non-resonant production of Z bosons in association with a photon.

Reducible background

- inclusive Z boson production in which a hadronic jet is misidentified as a photon



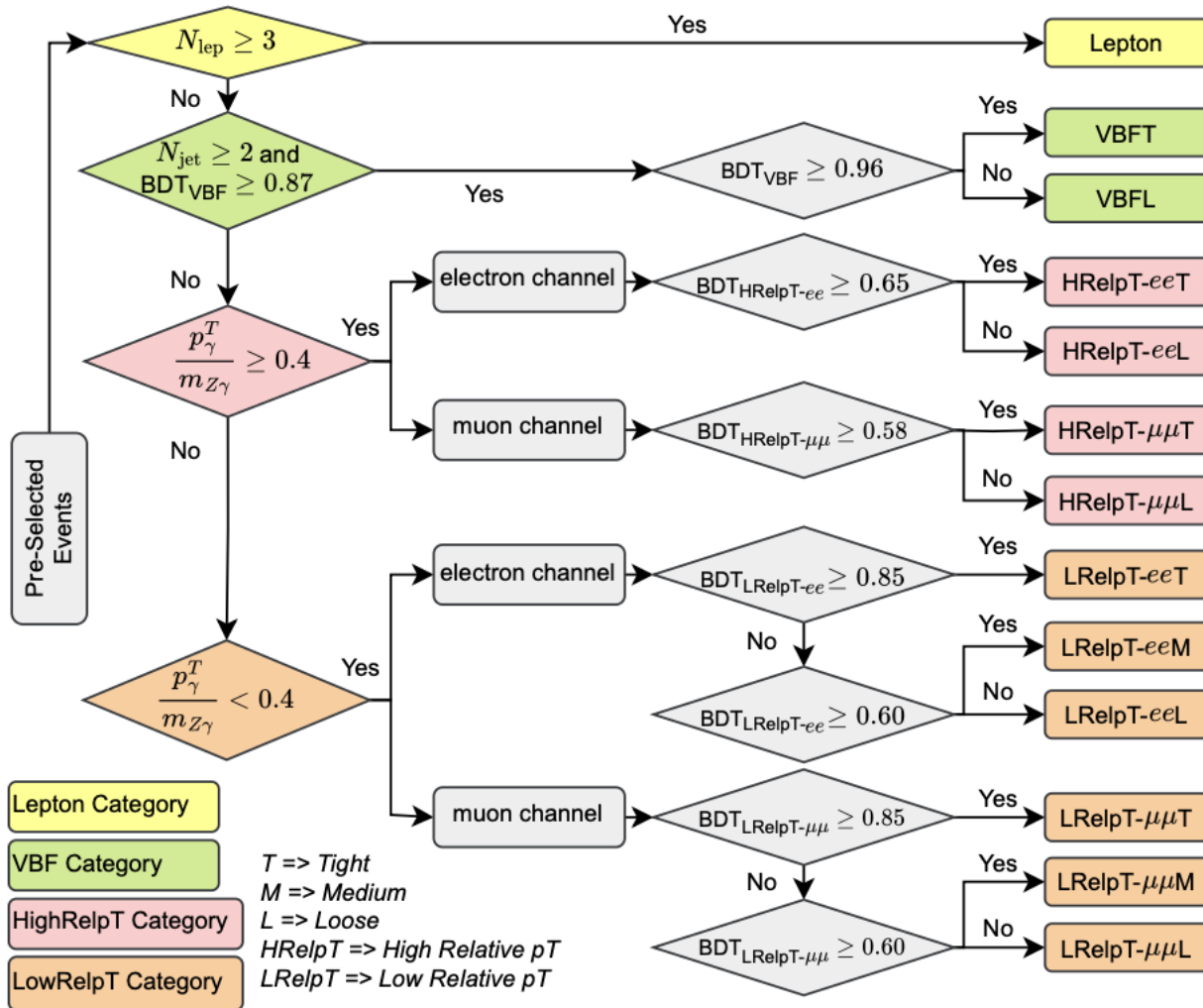
Event display of  $H \rightarrow Z\gamma \rightarrow ee\gamma$



Event display of  $H \rightarrow Z\gamma \rightarrow \mu\mu\gamma$



# categories



Events are classified into **13 categories**.

**Lepton category:** Additional leptons

**VBF category :** at least 2 jets

**High/Low Rel- $pT$  category :**

Separation between  $ee\gamma/\mu\mu\gamma$  final states

BDT classifications for VBF, HRelpT and LRelpT categories.

Split events by their BDT score into **Tight**, **(Medium)** **Loose** regions.



# Fit range optimization

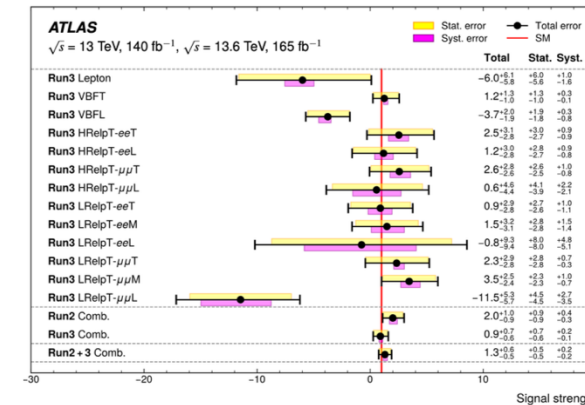
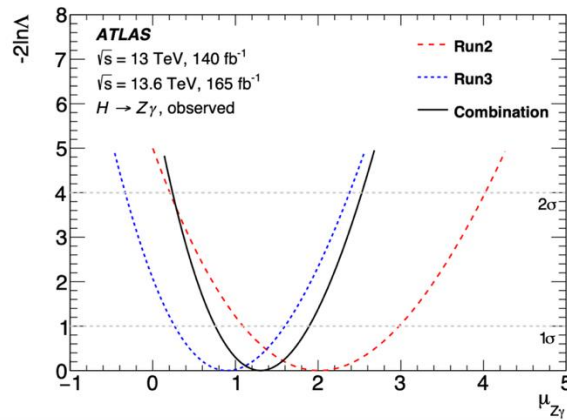
- In each category we build signal modeling and background modeling
- A larger fitting range will result in the best function tending to be of higher order and a larger spurious signal. Therefore, we adjust fit range depend on each category distribution.
- Calculated the results for multiple mass ranges to find the optimal range for the background model.

Mass range	Pvalue for same mu case	significance	DNLL
110-150	0.0326985	1.84253	3.39493
111-150	0.0339938	1.82509	3.33095
113-150	0.032083	1.85102	3.42629
114-150	0.0319835	1.85241	3.43142
115-140	0.0249291	1.96118	3.84622
115-150	0.0284554	1.904	3.6252



# Statistic result

Finally statistical results from the un-binned likelihood fit:



- Run 3 results:

measured signal strength  $\mu = 0.9^{+0.7}_{-0.6} \text{ (stat)} \pm_{-0.1}^{+0.2} \text{ (syst)}$

Run 3 observed (expected) significance:  $1.4 \sigma$  ( $1.5 \sigma$ ).

- Run 2 + Run 3 combination:

signal strength  $\mu = 1.3 \pm 0.5 \text{ (stat)} \pm 0.2 \text{ (syst)}$

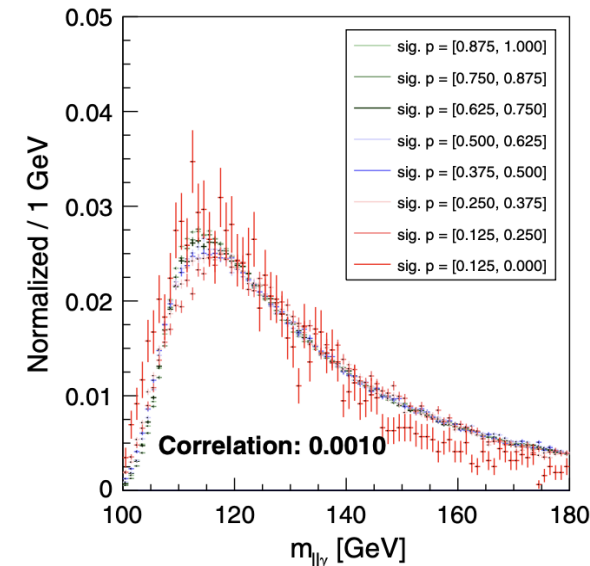
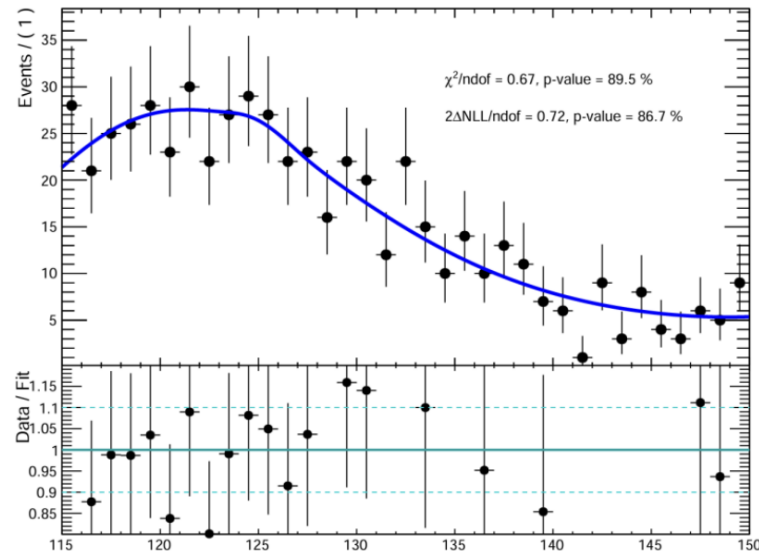
Run 2 + Run 3 combined observed (expected) significance:  $2.5 \sigma$  ( $1.9 \sigma$ ).

arXiv:2507.12598 (2025)



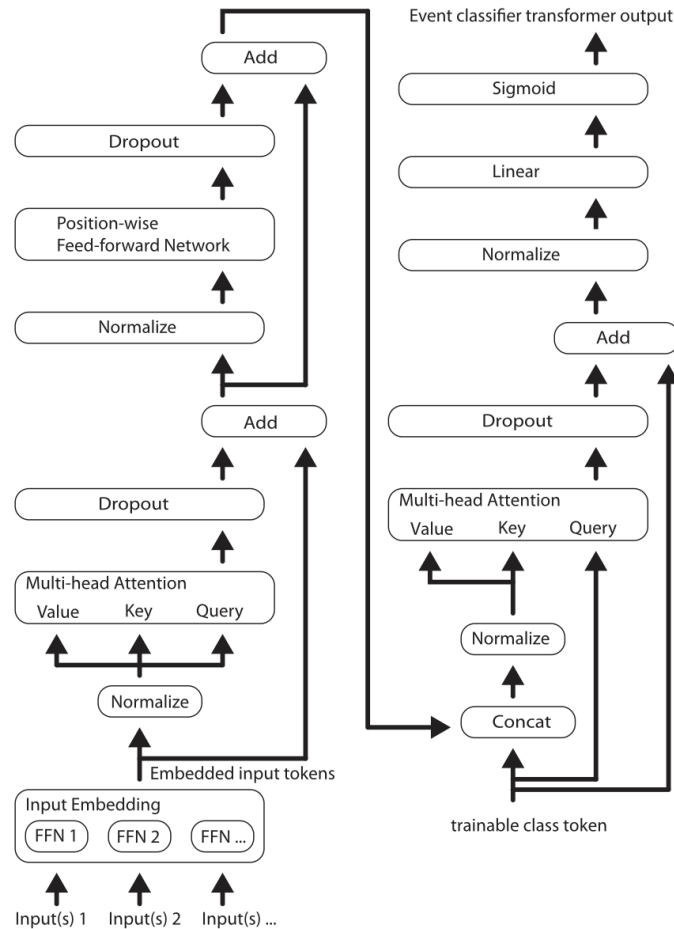
# Motivation of New Classifier

- In partial Run3 analysis, categorized by BDT shows high correlation with mass, which leads to difficulty in background modeling and spurious signal test.
- Recently study([kim2024PRD](#)) shows that a new transformer neural network with special loss function based on CMS H $\rightarrow$ Z $\gamma$  MC dataset can enhance the significance and reduce the correlation between the network's output and the reconstructed mass.
- The purpose of this work is to test and evaluate the feasibility of the new event classifier on Atlas H $\rightarrow$ Z $\gamma$  data, improve and use it in full Run3 H $\rightarrow$ Z $\gamma$  analysis.





- Event classifier transformer neural network



## Specialized loss function

Distance correlation (Disco) measures the dependence between output and mass

The DisCo term penalizes the neural network when  $y$  and mass are correlated.

$$\text{dCov}^2(X, Y) = \int d^p s d^q t |f_{X,Y}(s, t) - f_X(s)f_Y(t)|^2 w(s, t),$$

$$\text{dCorr}^2(X, Y) = \frac{\text{dCov}^2(X, Y)}{\text{dCov}(X, X)\text{dCov}(Y, Y)}$$

$$\text{Loss} = \text{Loss}_{\text{classifier}}(\hat{y}, y) + \lambda \cdot \text{DisCo}(\text{mass}, \hat{y}).$$

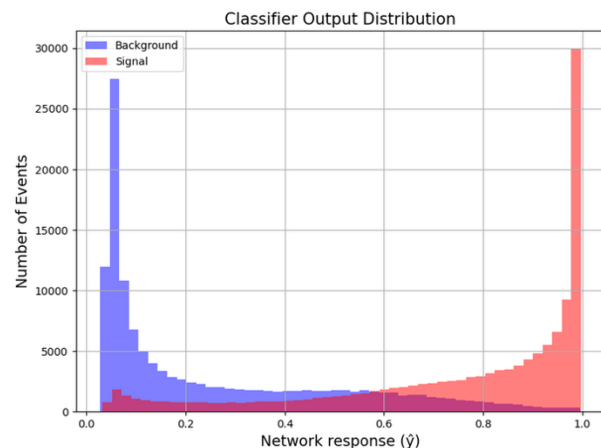


# Result of VBF training set

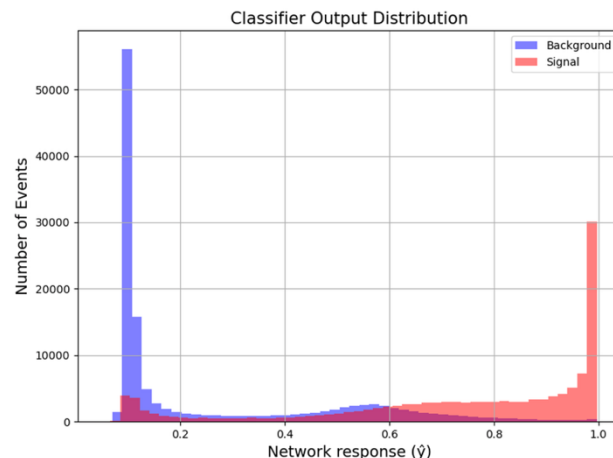


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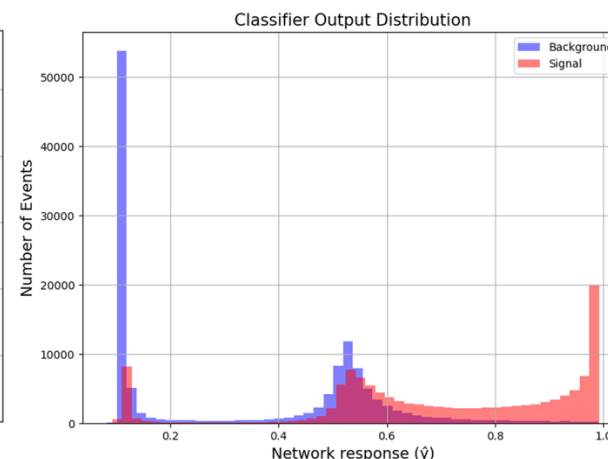
BCE + 10 DisCo



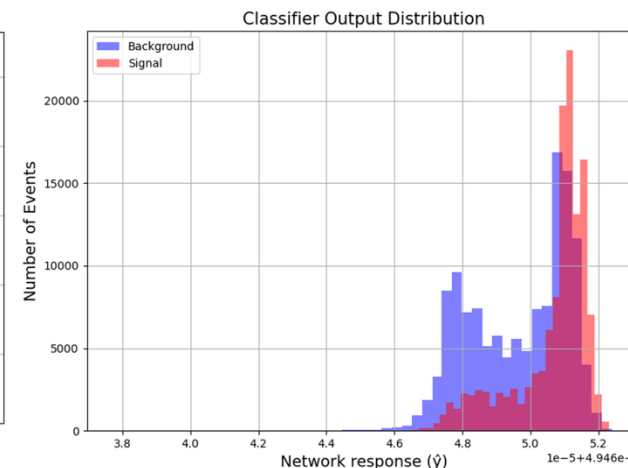
BCE + 15 DisCo



BCE + 20 DisCo



BCE + 30 DisCo



Factor of disco	10	15	20	30
avg loss	0.481855	0.512391	0.584306	0.856971
Significance	46.635 +- 8.847	42.973 +- 4.999	36.411 +- 3.738	23.782 +-5.549

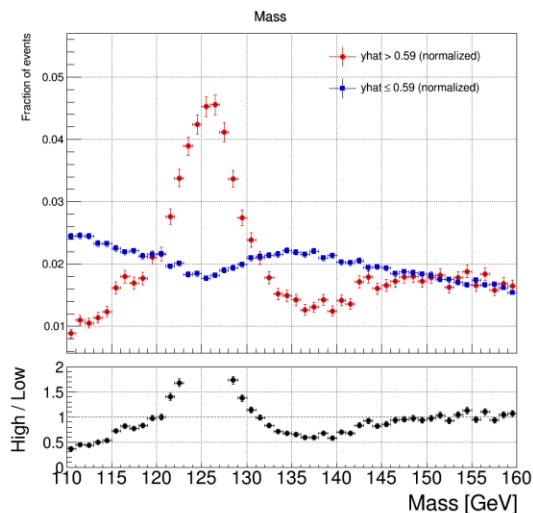


# Result of VBF training set

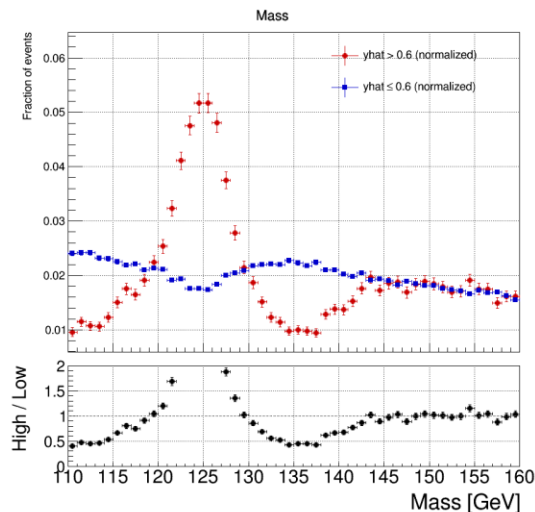


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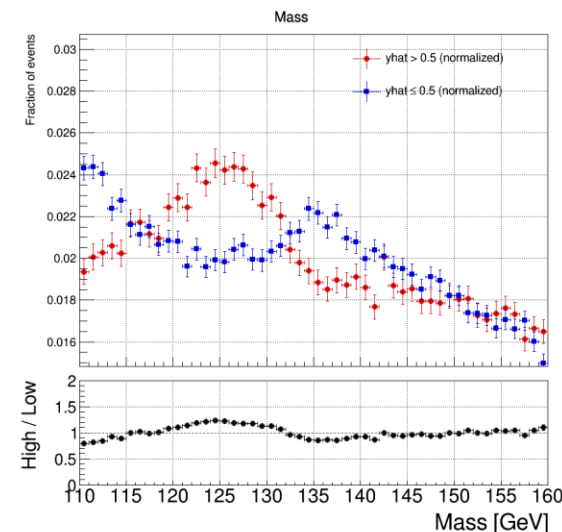
BCE + 10 DisCo



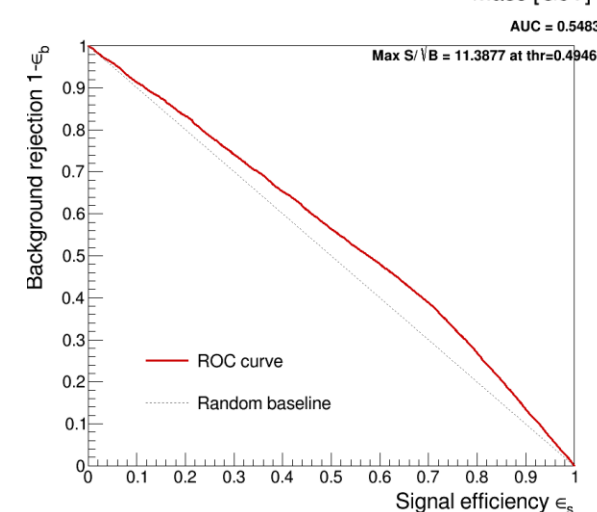
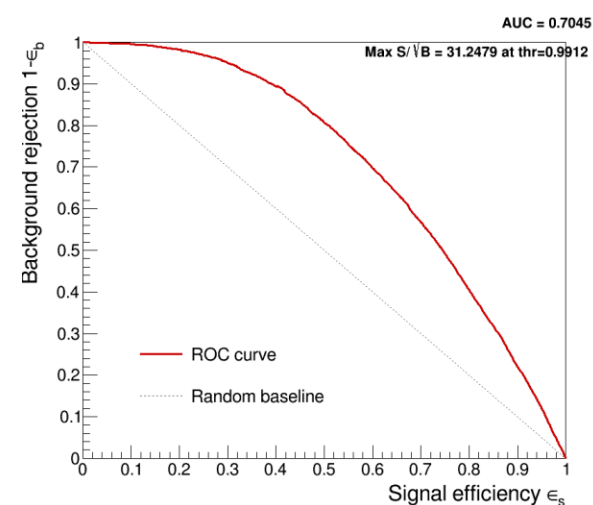
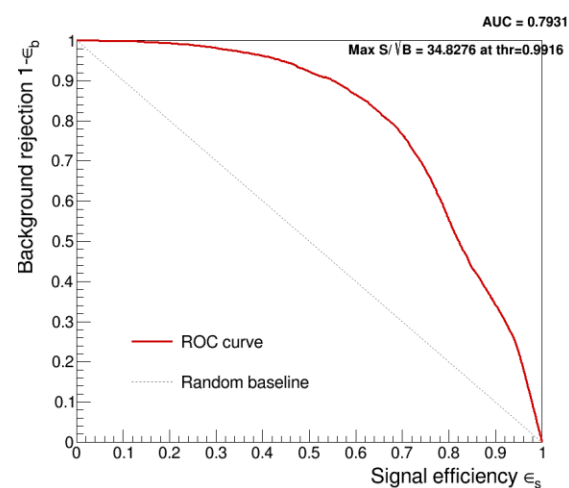
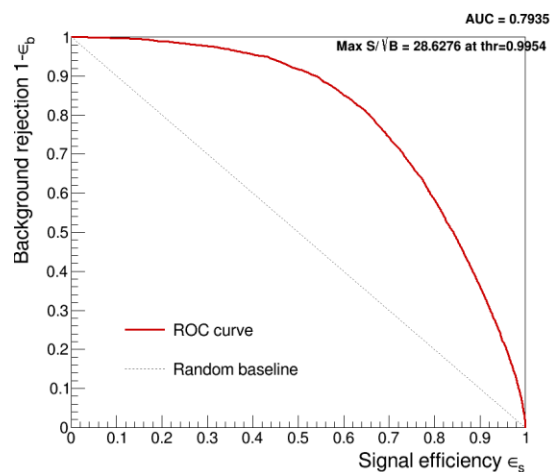
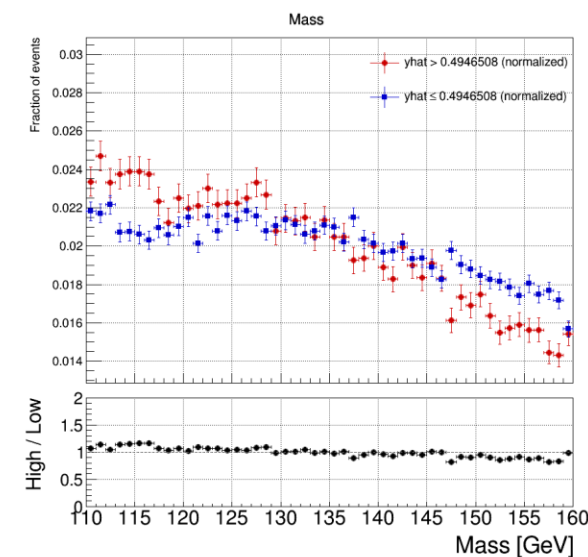
BCE + 15 DisCo



BCE + 20 DisCo



BCE + 30 DisCo





# Summary

- The Disco loss helps reduce the strong correlation with mass, but it also weakens classifier performance.
- I have applied the classifier transformer method and obtained preliminary results, next step will be to optimize the network parameters further and better understand its behavior.
- partial Run 3 HZy analysis.  
statistical work signal background model cross-check
- plan to the Full Run 3 analysis:  
event classifier, the signal and background models, statistical analysis.

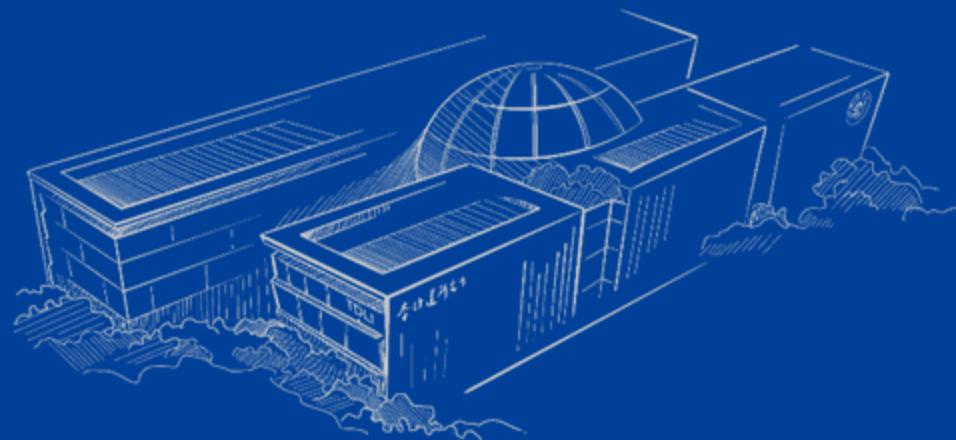
Thank you for your attention!





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# Backup





# Setup



- **Input** : x:features (25) y:classID (signal1 background0)
  - **Spectators**: llg\_m\_Zmassconstraint (for Disco)
  - **Model**:event classifier transformer neural network
  - **Loss function** :BCE +  $\lambda$  DisCo
- $$\text{BCE}(\hat{y}, y) = -y \ln(\hat{y}) - (1 - y) \ln(1 - \hat{y}),$$
- **output**: probability of the event (sigmoid)
  - Batchsize 1000 epoch 1000

**Key factor**: loss, significance, AUC of ROC

**Model selection**: Significance-best

- Divide the dataset into 5 bins based on output. The bins are constructed to have an equal number of signal events. Calculate the significance of each bin and combine the significances of the bins.

$$\text{Significance} = \sqrt{2 \left[ (N_S + N_B) \ln \left( 1 + \frac{N_S}{N_B} \right) - N_S \right]},$$

$$\text{Total significance} = \sqrt{\sum_i^n (\text{Significance}_i)^2}$$

```
# VBF_DRmin_y_j
# VBF_Dphi_Zy_jj
# VBF_Dy_j_j
# VBF_N_j
# VBF_Zepp
# VBF_eta_j1
# VBF_eta_j2
# VBF_m_jj
# VBF_mass_j1
# VBF_mass_j2
# VBF_pT_j1
# VBF_pT_j2
# VBF_pT_jj
# VBF_pTt_Zy
# Zy_Dphi_j1
# ll_eta
# ll_pt
# llg_angles_cotheta_ginH
# llg_angles_cotheta_linZ
# llg_deta_Zy
# llg_dphi_Zy
# llg_eta
# llg_pt
# ph_eta
# ph_pt
```

Variables	Definition
$M_{jj}$	Di-jet mass
$\Delta\eta_{jj}$	Pseudo-rapidity separation of dijet
$pT_{j1}$	lead-jet pT
$M_{j1}$	lead-jet mass
$\eta_{Zeppenfeld}$	$ \eta_{Z\gamma} - 0.5 * (\eta_{j1} + \eta_{j2}) $
$\eta_{j2}$	sublead-jet eta
$pT_{jj}$	Di-jet pT
$\Delta\Phi_{Z\gamma,jj}$	Azimuthal angle between Zgamma and dijet system
$M_{j2}$	sublead-jet mass
$\Delta R_{\gamma or Z,j}^{min}$	Minimum $\Delta R$ between one object of the Zgamma and jets
$pT_{j2}$	sublead-jet pT
$\eta_{j1}$	lead-jet eta
$pT_{l\gamma}$	$ll\gamma$ pT
$\Delta\Phi_{Z\gamma,j1}$	Azimuthal angle between Zgamma and lead-jet
$pT_{\gamma}$	photon pT
$pT_{ll}$	$ll$ pT
$N_j$	Number of jets pT
$\Delta\Phi_{Z,\gamma}$	Azimuthal angle between di-lepton system and photon
$\cos\theta(ll,\gamma)_{inZ}$	$\cos\theta$ in Z rest system
$\eta_{ll}$	eta of Z
$\eta_{ll\gamma}$	eta of $ll\gamma$
$\cos\theta(ll,\gamma)_{inH}$	$\cos\theta$ in Higgs rest system
$\eta_{\gamma}$	eta of photon
$pT_{\gamma}$	Zgamma $p_T$ projected perpendicular to the Zgamma thrust axis
$\Delta\eta_{Z,\gamma}$	Pseudo-rapidity separation of Z $\gamma$

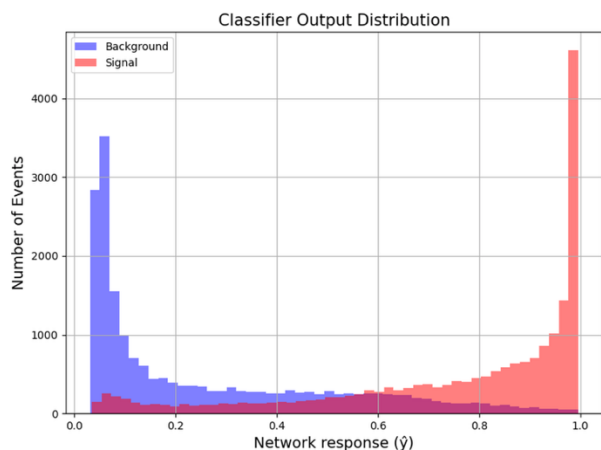


# Result of validation set

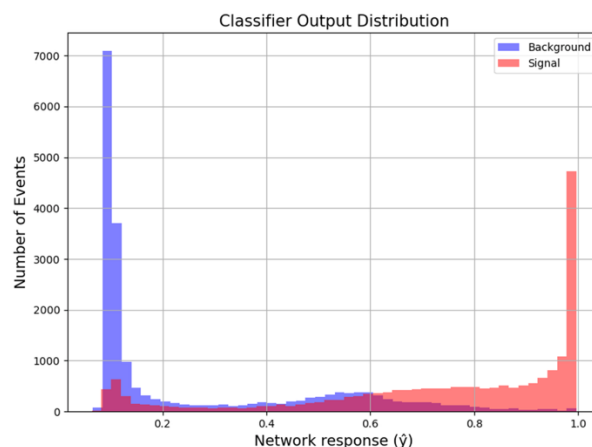


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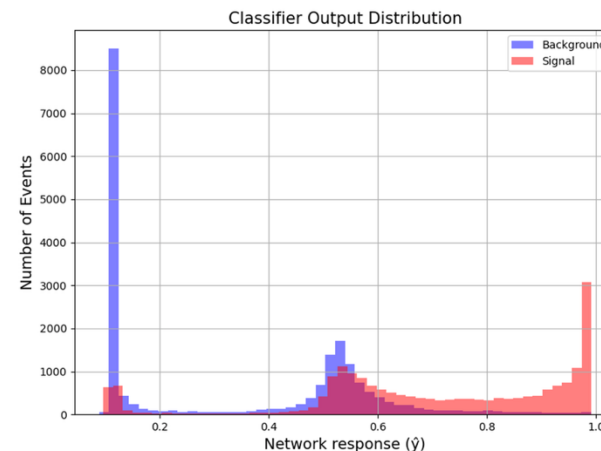
BCE + 10 DisCo



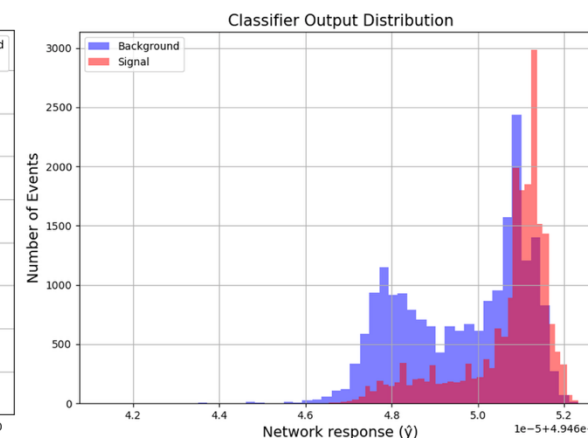
BCE + 15 DisCo



BCE + 20 DisCo



BCE + 30 DisCo



Factor of disco	10	15	20	30
avg loss	0.500804	0.516626	0.585283	0.868690
Significance	13.923 +- 1.817	13.648 +-1.817	14.397 +- 2.479	9.428 +- 0.616

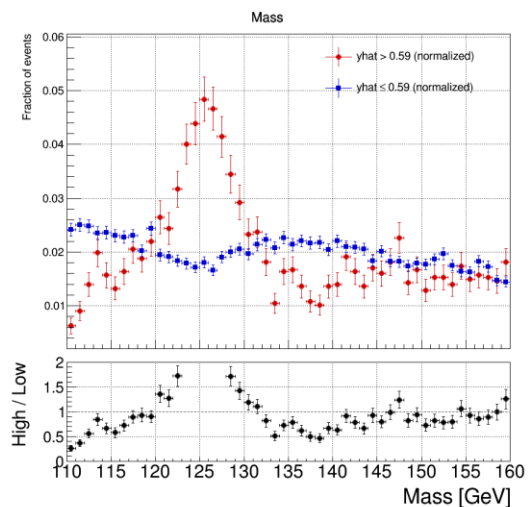


# Result of validation set

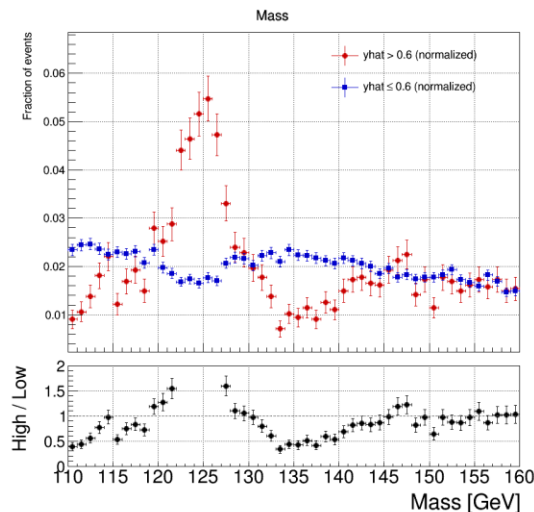


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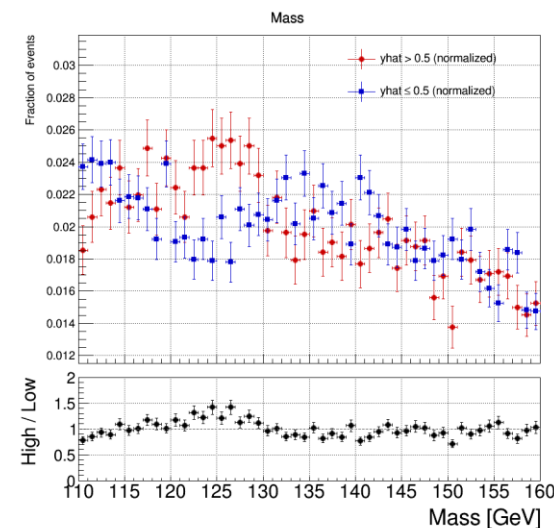
BCE + 10 DisCo



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