

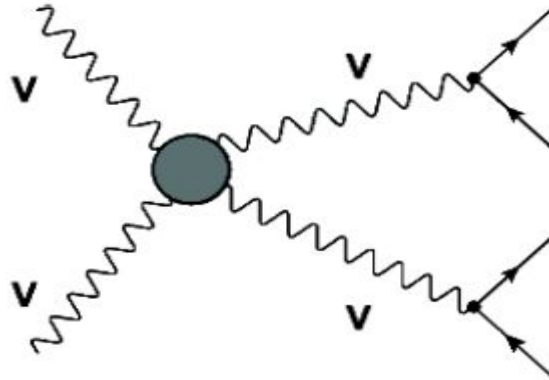
The Search for Evidence of Vector Boson Scattering between a W boson and a Photon in proton-proton collisions

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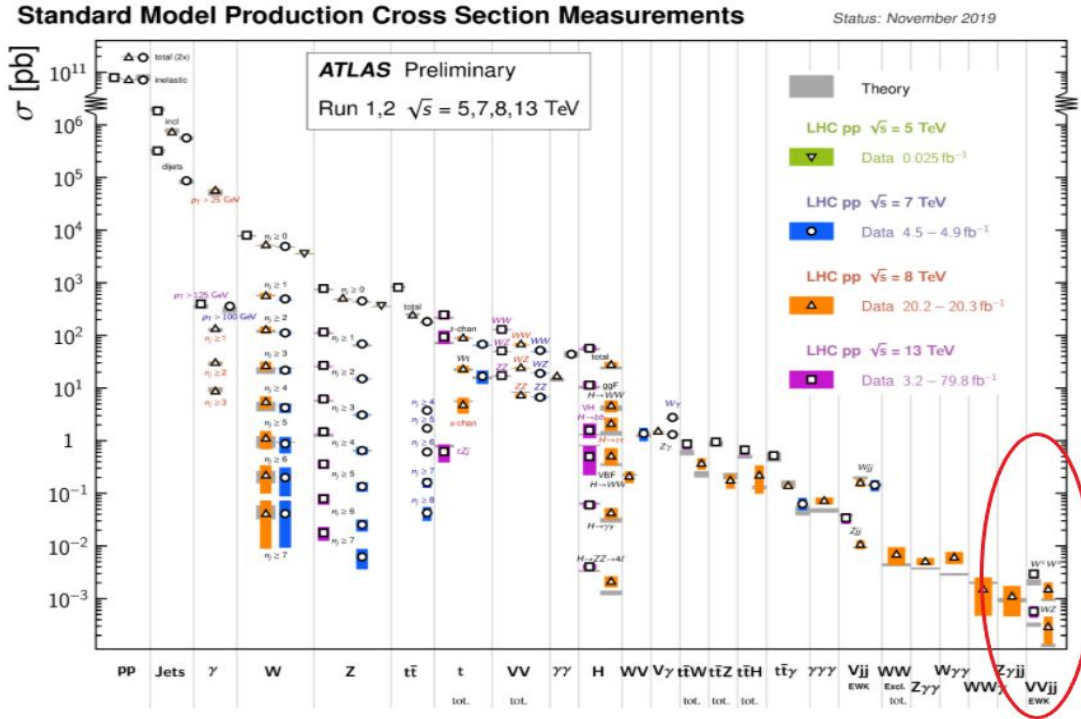


Why Study Vector Boson Scattering (VBS)?

- In the Standard Model, the interactions between gauge bosons are completely specified by the $SU(2) \times U(1)$ structure of the theory.
- This makes the study of the interactions between gauge bosons a powerful approach to search for new physics.
- Any deviation from SM predictions in the self interactions of gauge bosons would indicate the presence of new physics phenomena.



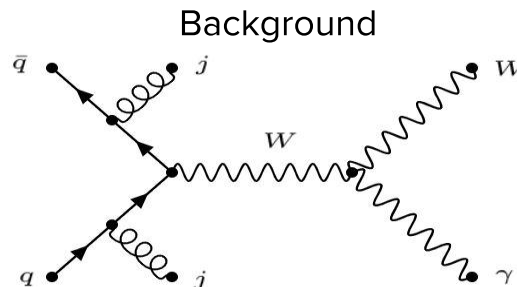
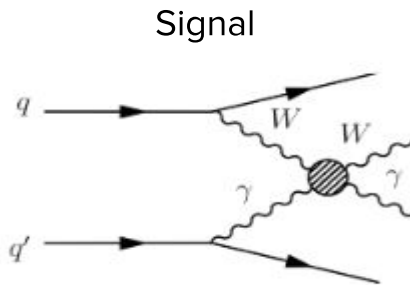
VBS at the LHC



- The LHC provides a unique environment in which to study rare Standard Model processes.
- The search for evidence of scattering between a W boson and a photon is carried out using a total of 139 fb⁻¹ data collected by the ATLAS detector at $\sqrt{s} = 13$ TeV.

Challenges of VBS Measurements

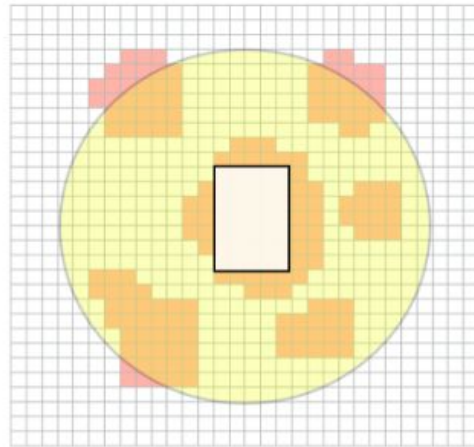
- The search for the scattering of a W boson and a photon comes with formidable challenges.
 - There is a large and irreducible background from poorly modelled processes involving strong interactions between particles (QCD):



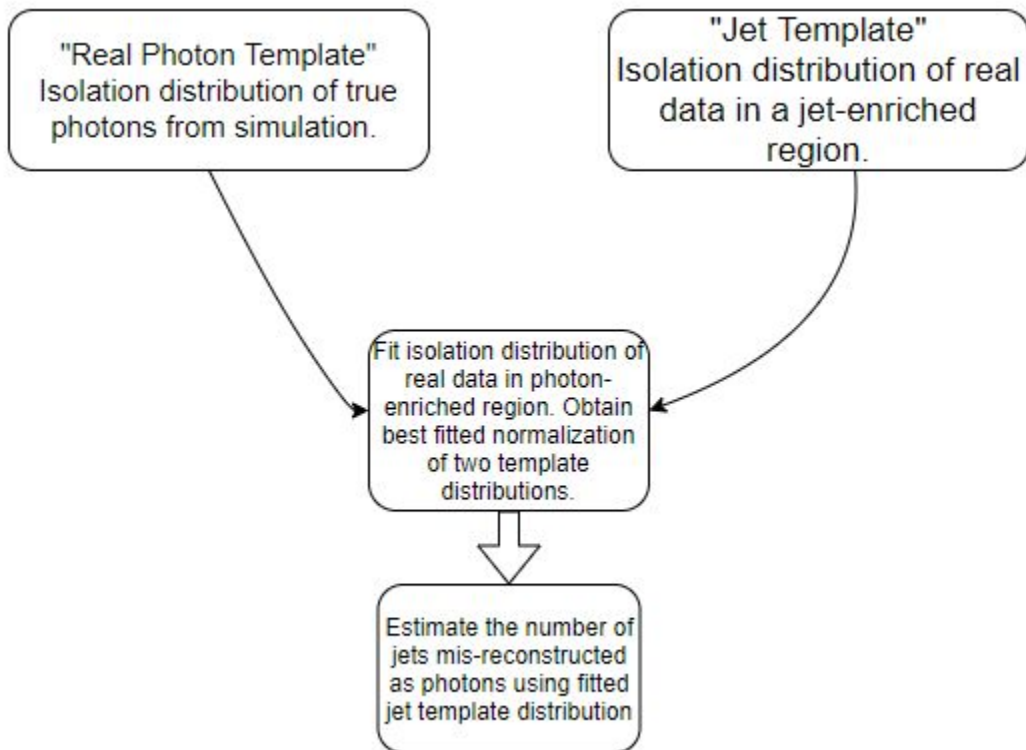
- The imperfect modelling of the detector response results in a non-negligible number of jets being misidentified as photons.
- In this talk I will discuss a **data driven approach to estimating the number of jets misidentified as photons** and a **machine learning approach to estimating the number of signal events from the large irreducible background**.

Jets Faking Photons - A Data Driven Approach

- Simulated events cannot be used to estimate the size of this background due to the imperfect modelling of the detector response to jets.
- Photon Identification relies on measurements in the calorimeter.
 - A photon candidate satisfying various cuts on calorimeter variables is considered tight.
 - A photon candidate that fails a subset of these cuts is considered non-tight.
 - A tight and isolated photon candidate is considered a signal photon.
- Photon candidate identification variables are uncorrelated with isolation energy
 - The shape of the isolation distribution for tight and non-tight photon candidates, originating from jets, is the same.
 - Use this property to estimate the number of jets misidentified as photons.

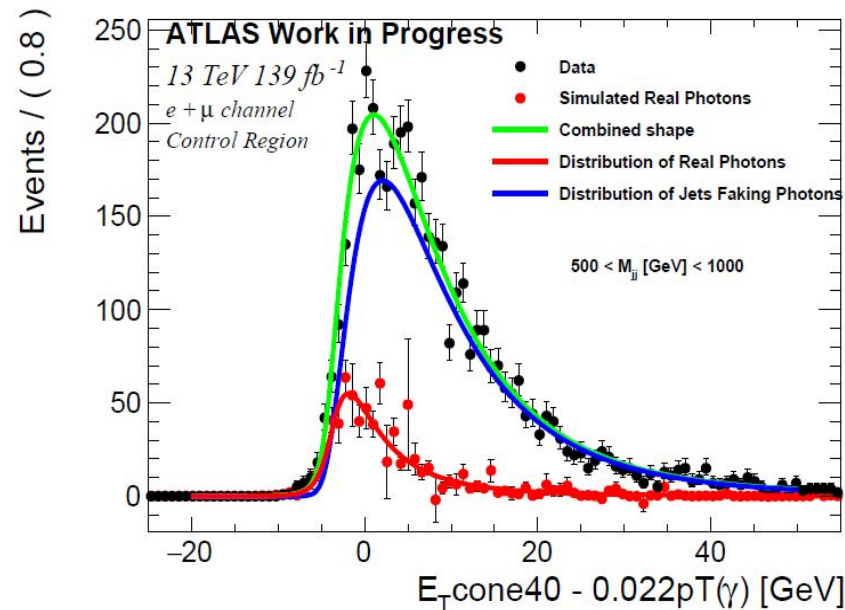


The Template Fake Method



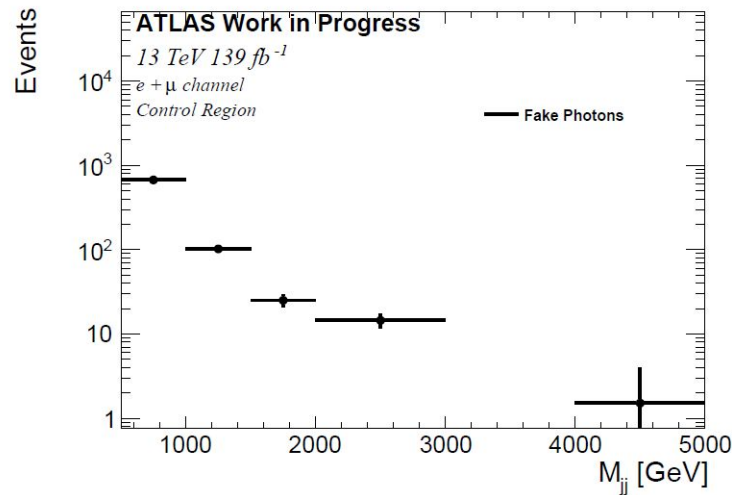
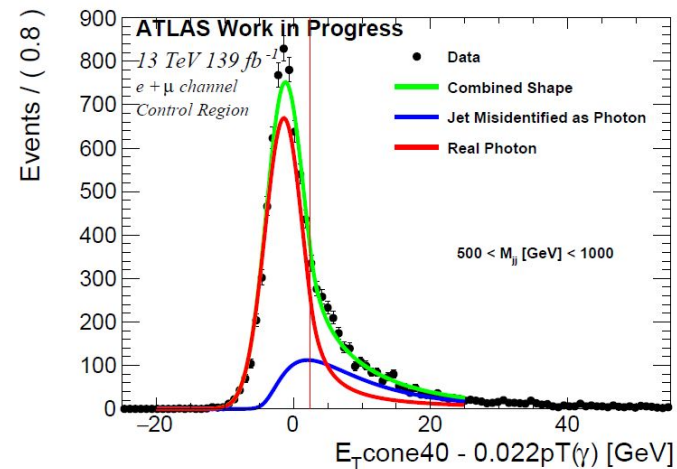
Determining the isolation shape of jets mis-identified as photons

- Jet enriched region still contains a small fraction of true photons.
 - Determine the shape of the isolation of true photons from simulated data.
 - Assume a functional form for jets contribution.
 - Fit combined shapes to real data.
- Shape of isolation of jets mis-identified as photons given by best-fit parameters of jet function (blue curve).



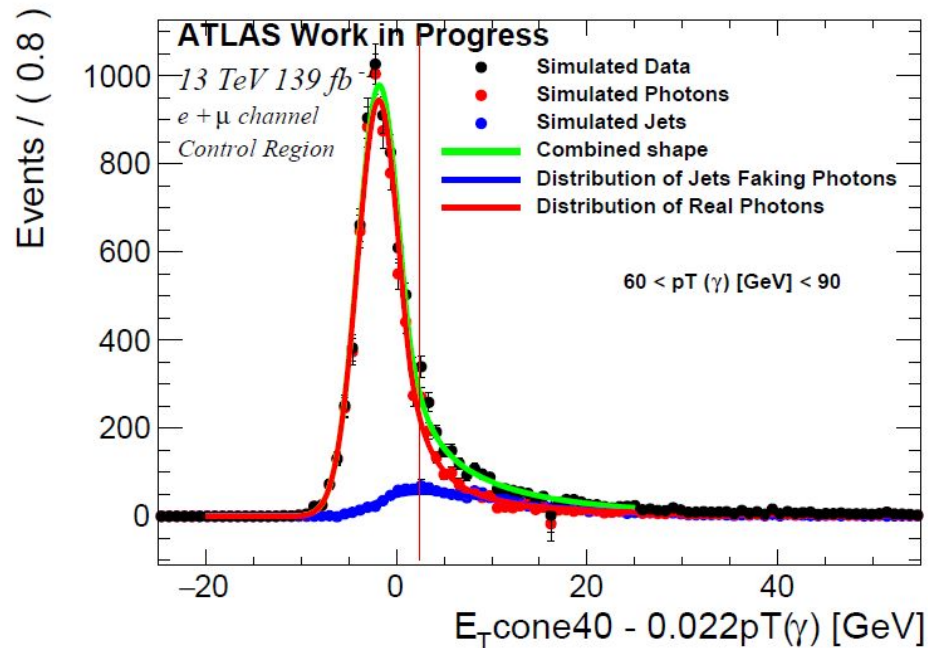
Determining the number of reconstructed photons originating from mis-identified jets

- Photon-enriched region contains a small fraction of jets.
 - Use isolation shape of true photons from simulated data.
 - Use isolation shape of jet distribution determined in previous step.
 - Fit combined shapes to real data, taking normalizations as free parameters..
- Number of jets mis-reconstructed as photons obtained by integrating jet distribution in the signal region (-25 to 2.45).



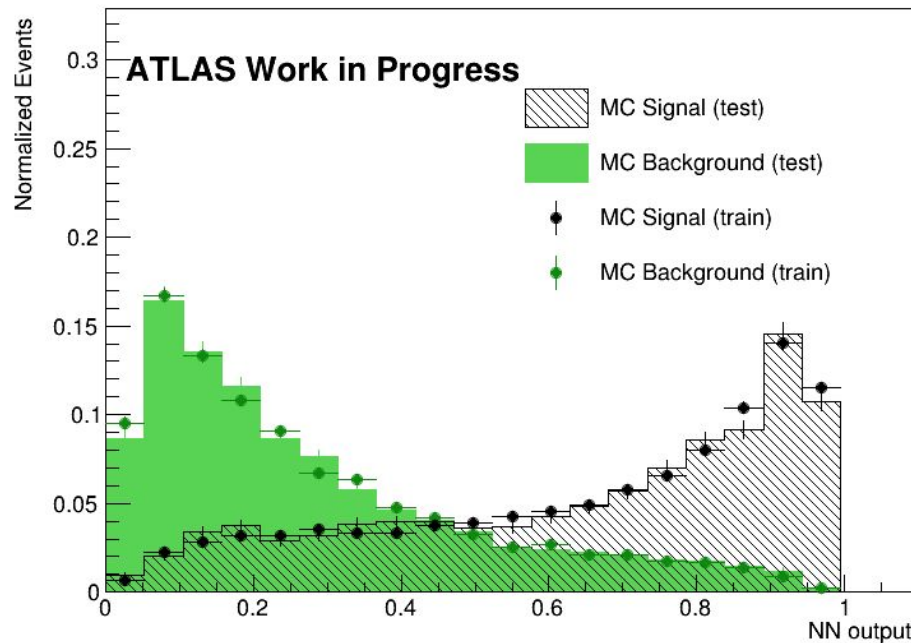
Checking the Method

- Verify method by creating a testing dataset from simulated data.
- Estimated distribution of jets mis-reconstructed as photons (blue line) matches simulated sample (blue points).



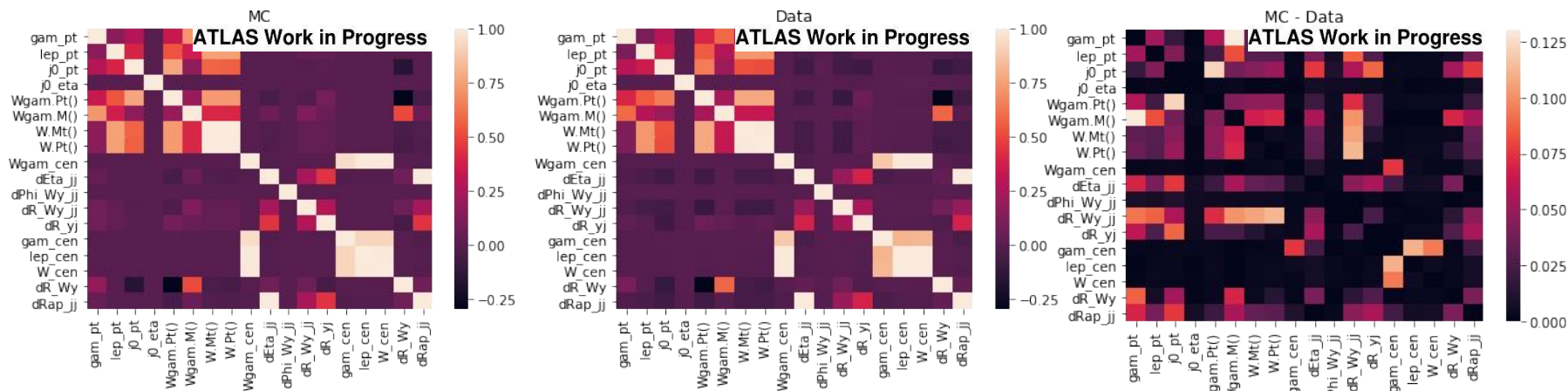
Distinguishing Electroweak Signal from Irreducible Background

- Large irreducible background from QCD processes.
- To leverage the discriminating power of multiple variables, use them to train a neural network.
 - 5 layer fully connected neural network.
 - Trained on 18 variables.
 - 60% of simulated sample used as training set, 20% used as testing set.
- NN distinguishes signal from QCD background with 69% accuracy.
 - Signal defined as event with NN score > 0.5 .



Generalizing performance from MC to data

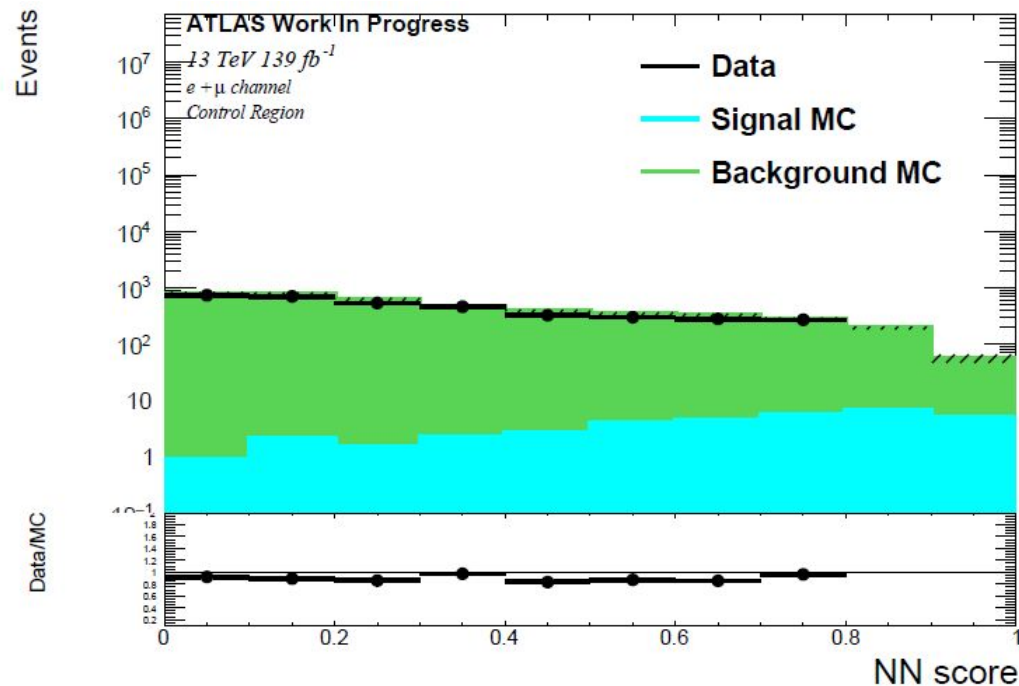
- The neural network will learn correlations between training variables. Similar performance can be expected when evaluating the model on data if the correlation matrices are similar.



- Maximum difference of .12 suggests performance on data will be similar to performance on simulated data.

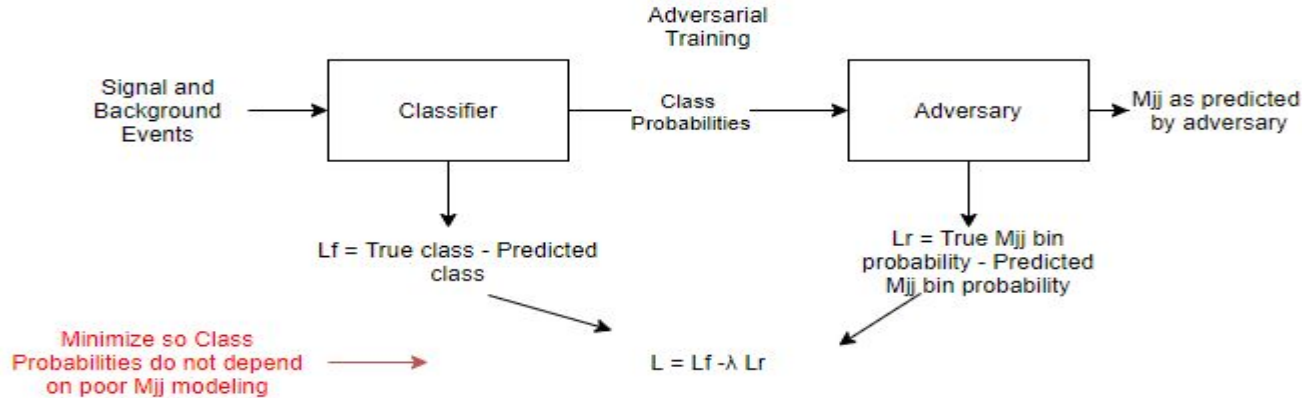
Comparing Simulated and Real Data NN output

- Apply NN to real data in background-enriched control region to verify agreement between data and simulation.
- NN model generalizes well onto real data.



Next Steps: Adversarial Training

- Irreducible QCD background is mismodelled at high values of di-jet invariant mass (M_{jj})
- Train a second neural network to learn the M_{jj} distribution (predict M_{jj} bins) from the output of the classifier.



Summary

- VBS measurements provide a powerful probe of new physics.
- The search for the vector boson scattering of a photon and a W boson comes with two significant challenges: a large background from jets misidentified as photons and a large irreducible QCD background.
- A data driven approach has been shown to effectively estimate the background from jets faking photons.
- A neural network has been trained to strongly discriminate the electroweak signal from the dominant QCD background.
- New techniques being explored for regularizing the neural network to not learn MC mismodelling.