

[ MACHINE LEARNING in the SEARCH FOR NEW PHYSICS PHENOMENA AT THE LHC ]

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Deepening the  
Distance Between  
Collider Events

Transferability of  
Deep Learning in  
Searches for  
New Physics

DL-assisted  
Strategies Towards  
Generic Searches  
for Exotic Particles

How transferable is a DL model to a new target signal, unknown to the model?

In computer vision, the first layers of DL models are highly transferable to other tasks. They learn about localised pixel variations, textures and patterns and only the last layers distinguish the high-level picture such as *dog or cat* [1]

We develop Deep Neural Networks (DNN) implemented with Keras using TensorFlow as backend to probe transferability in the HEP domain. DNN Bayesian optimisation from Scikit-Learn is used.

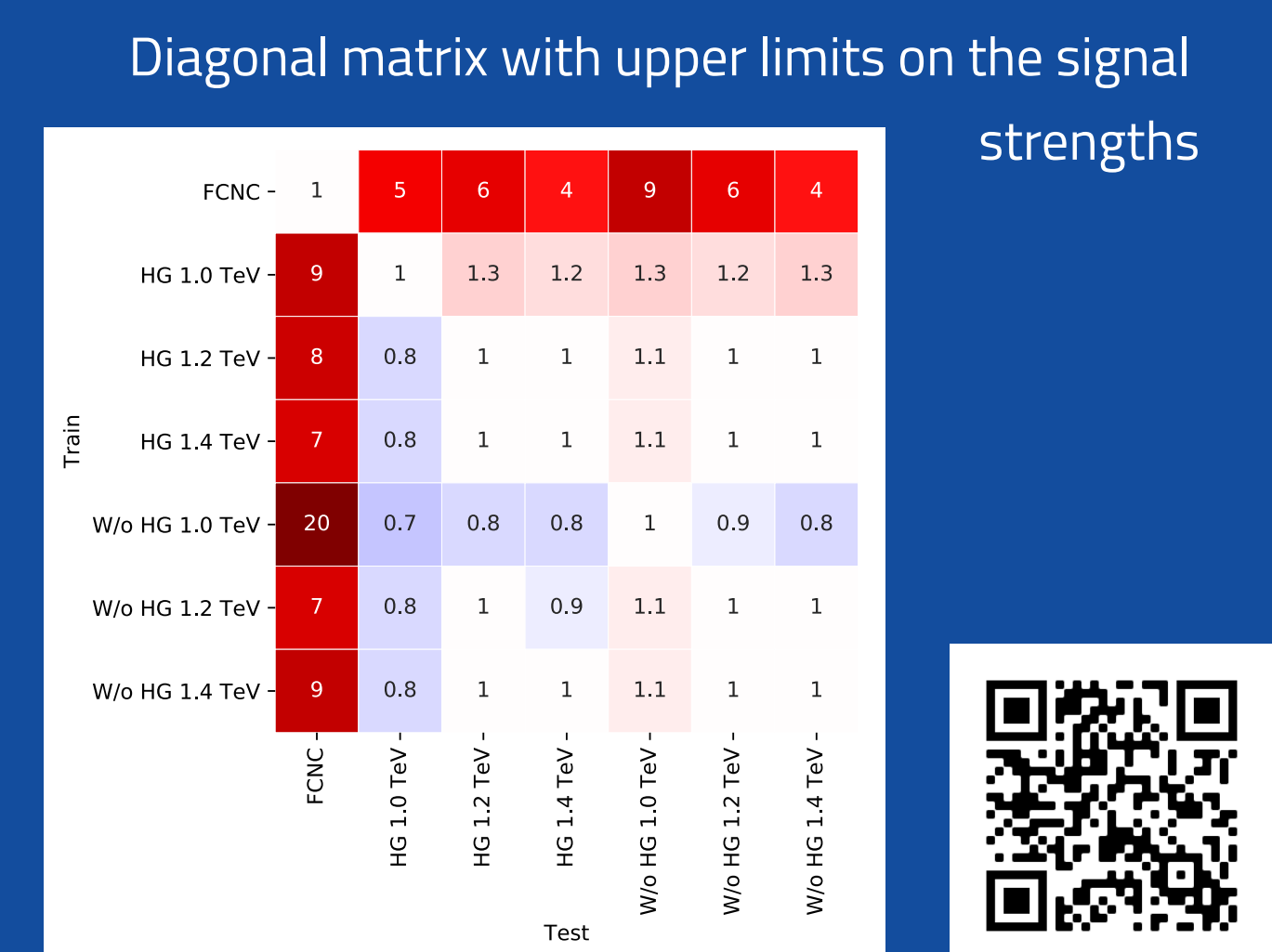
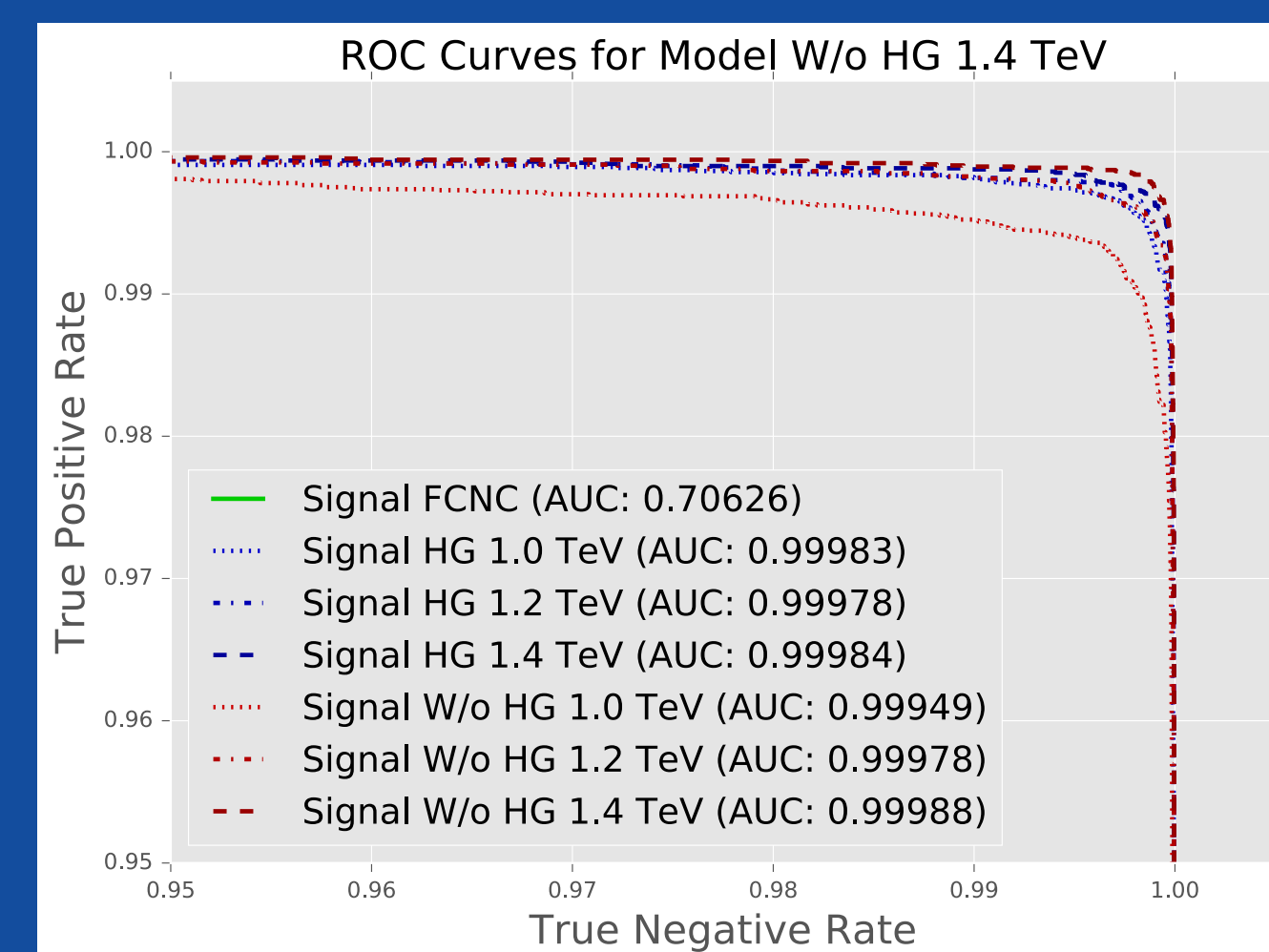
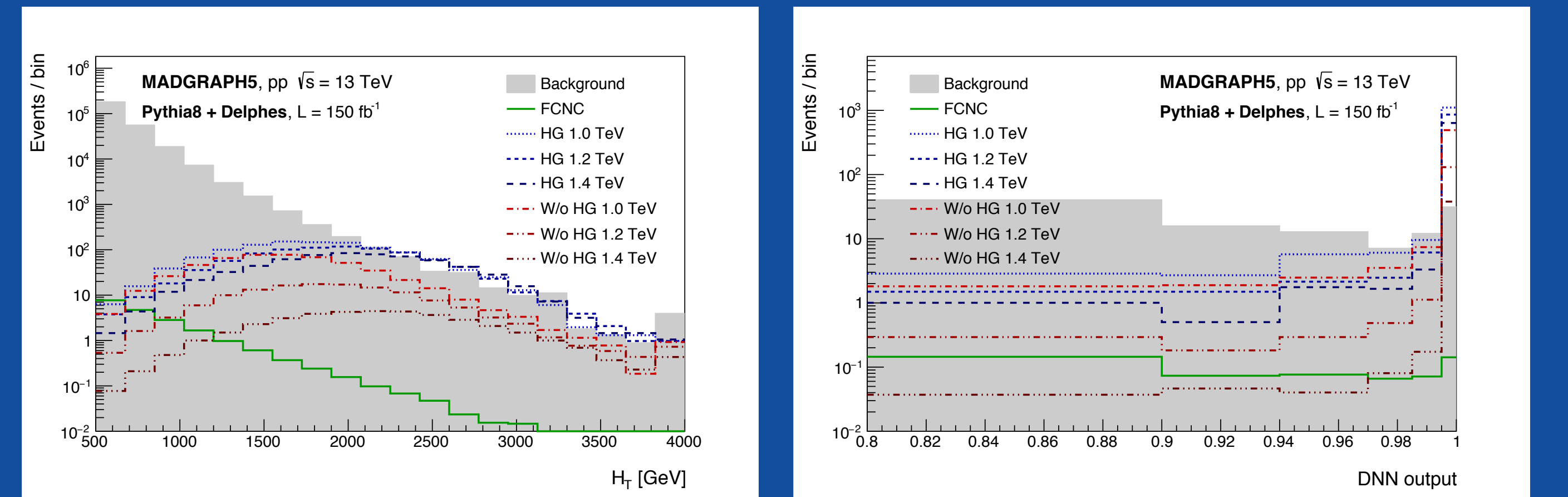
Benchmark signals contain  $t\bar{t}Z$  final states, and are simulated with MadGraph5+Delphes:

- $t\bar{t}Z$  production via FCNC vertex  $ut\bar{t}Z$
- Vector-Like Top pair production (TT) via SM gluon or BSM heavy gluon

DNN trained to classify each signal type and tested for each signal case:

- Model is able to identify other signal types as non-background
- Exclusion limits computed from DNN discriminant are nearly independent of the train signal

[1] J. Yosinski et al. How Transferable are features in deep neural networks?, 2014



Are rare events distant from more common ones?

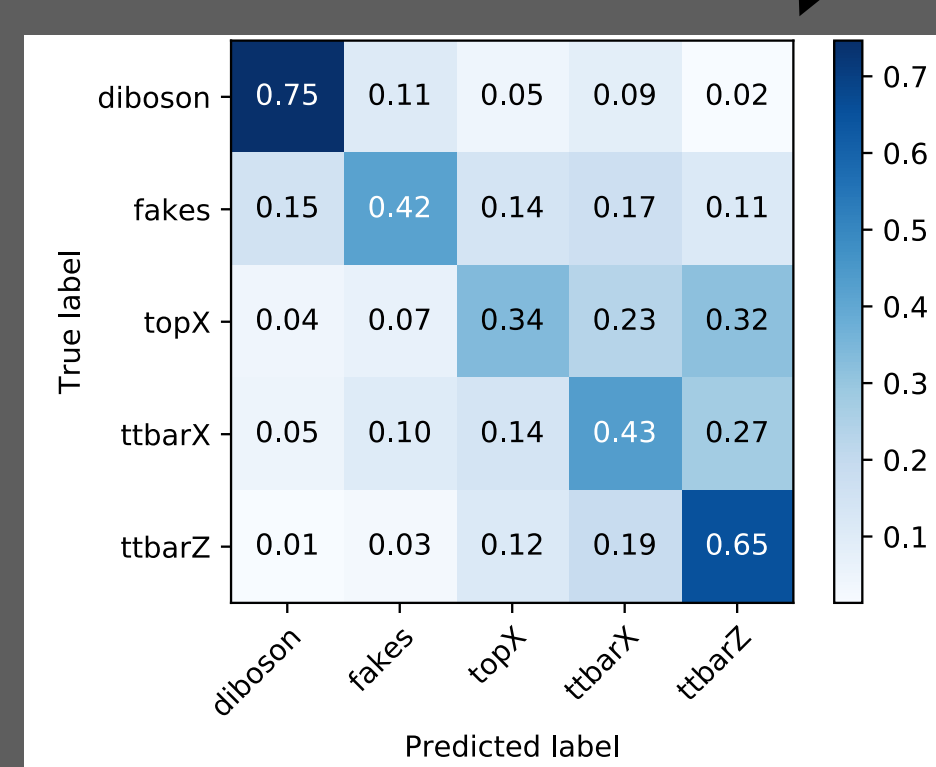
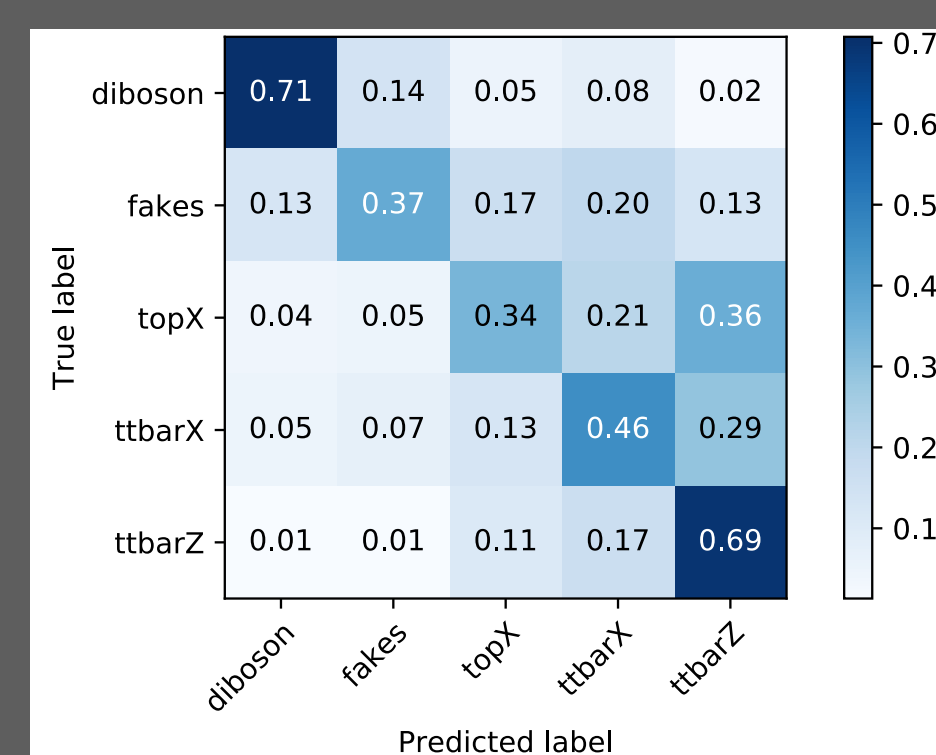
An event metric was recently introduced for the case of jet constituents, assisting in the distinction between boosted top quarks and QCD jets [2].

We deploy the idea of the event metric to fully reconstructed events:

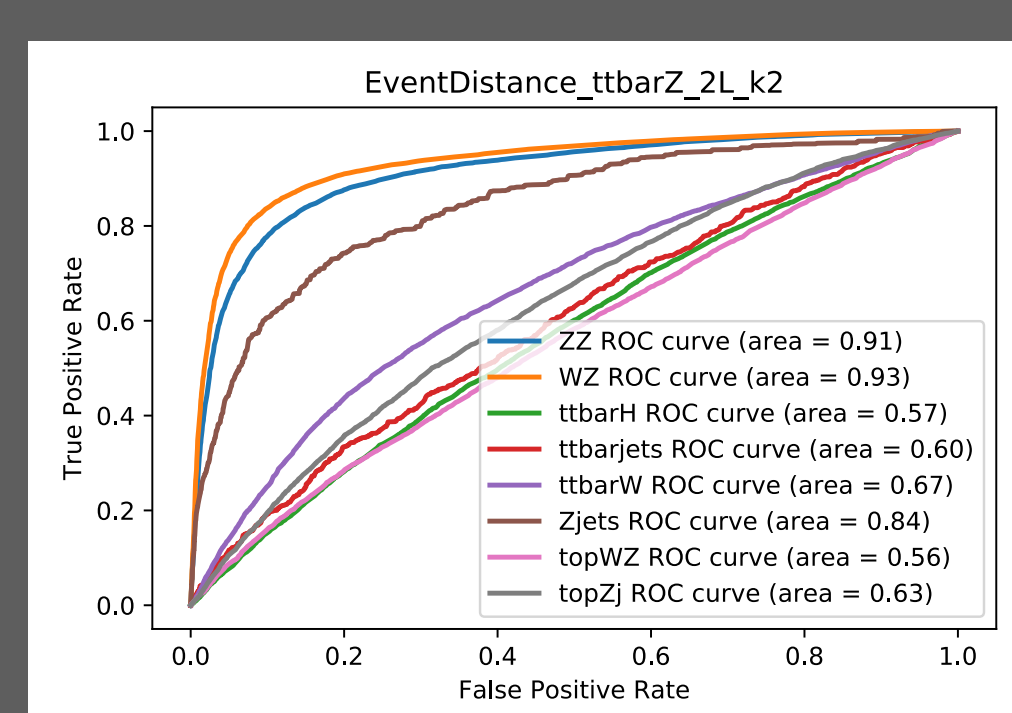
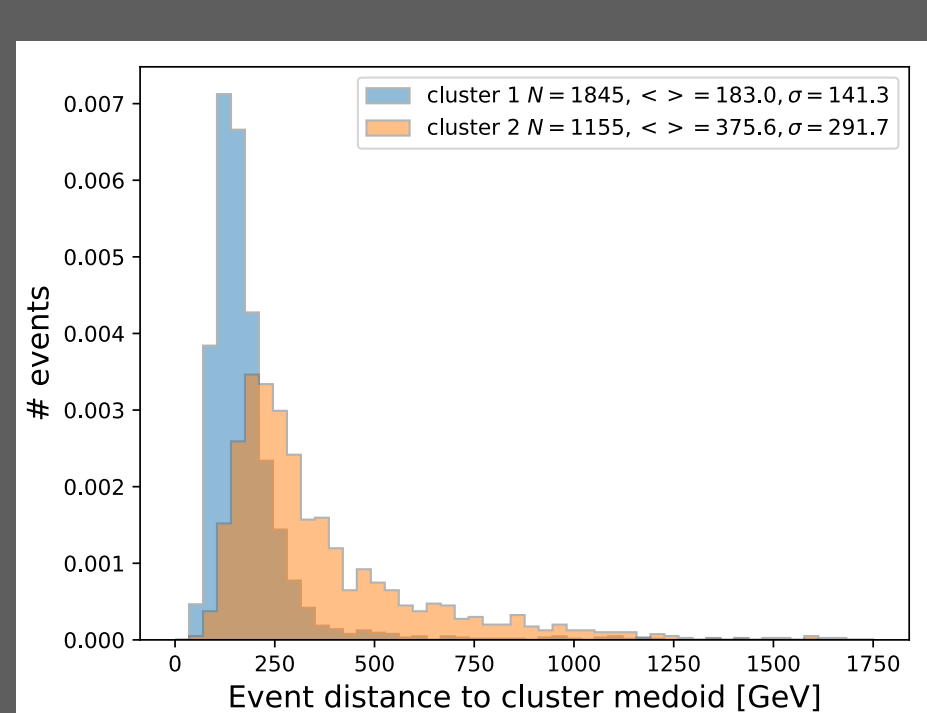
- Distance between events  $I$  and  $J$   
$$d(I, J) = \min_{i,j} (\sum \Delta R_{ij} \times |p_{T,i} - p_{T,j}| \times ID(i, j))$$
- Minimize the equation with the Earth Mover's Distance algorithm implemented in the Python Optimal Transport library
- Cluster close events and define the central medoid events as references per each process
- Compute the distance to the medoids to access how near/far the event is from the represented process

Study rare  $t\bar{t}Z$  production as physics case:

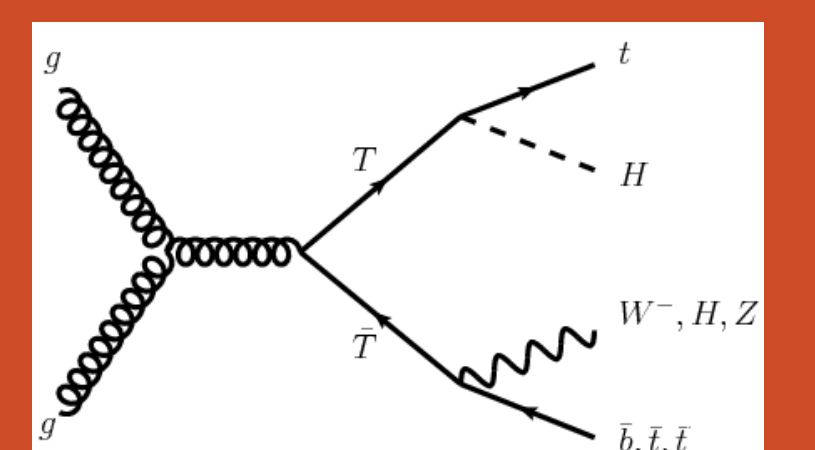
- Distance to  $t\bar{t}Z$  medoids discriminates against diboson and Z+jets
- Performance of a multi-class DNN improved with distance observables



[2] P. T. Komiske et al. The Metric Space of Collider Events, 2019



Which strategy to search generically for Vector-Like Quarks, regardless of the final state topology?



VLQs arise in BSM theories addressing the Naturalness problem with mechanisms cancelling the quadratically-divergent contributions to the Higgs mass.

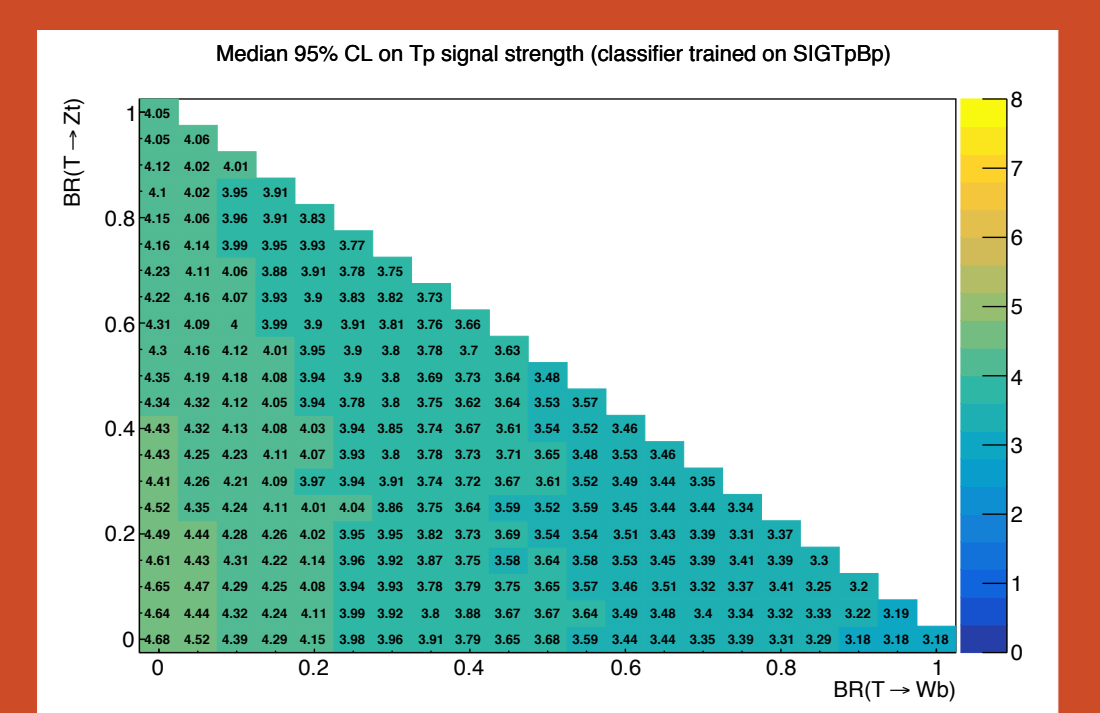
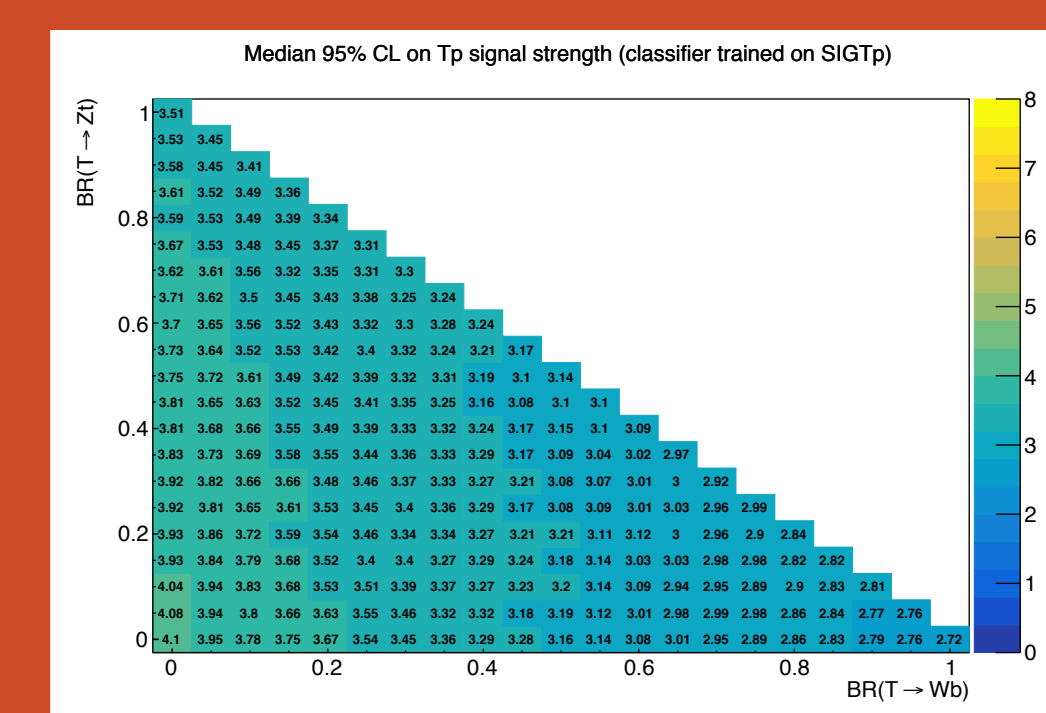
Leading to diverse final states, ATLAS, for instance, needs 17 analyses to reach maximal sensitivity to VLQs [3].

DNN trained with 4-momentum of final state objects:

- Pre-selecting events with at least 1 lepton and large energy deposits excludes most of the background: DNN will focus on difficult cases

Single model trained with signal at fixed branching ratio (BR):

- DNN discriminant has good sensitivity to the full BR plane
- Model trained with BB ~equally sensitive to the TT signal



[3] ATLAS Collaboration, Combination of the searches for pair-produced vector-like partners of the third-generation quarks at  $\sqrt{s}=13$  TeV with the ATLAS detector, 2018