

Higgs to WW - From Peter Higgs to Deep Learning and Histogram Rebinning

Benjamin Jäger

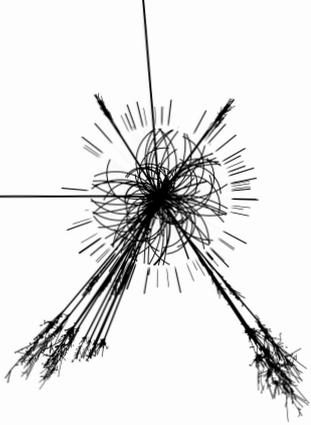
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WNPCC2020

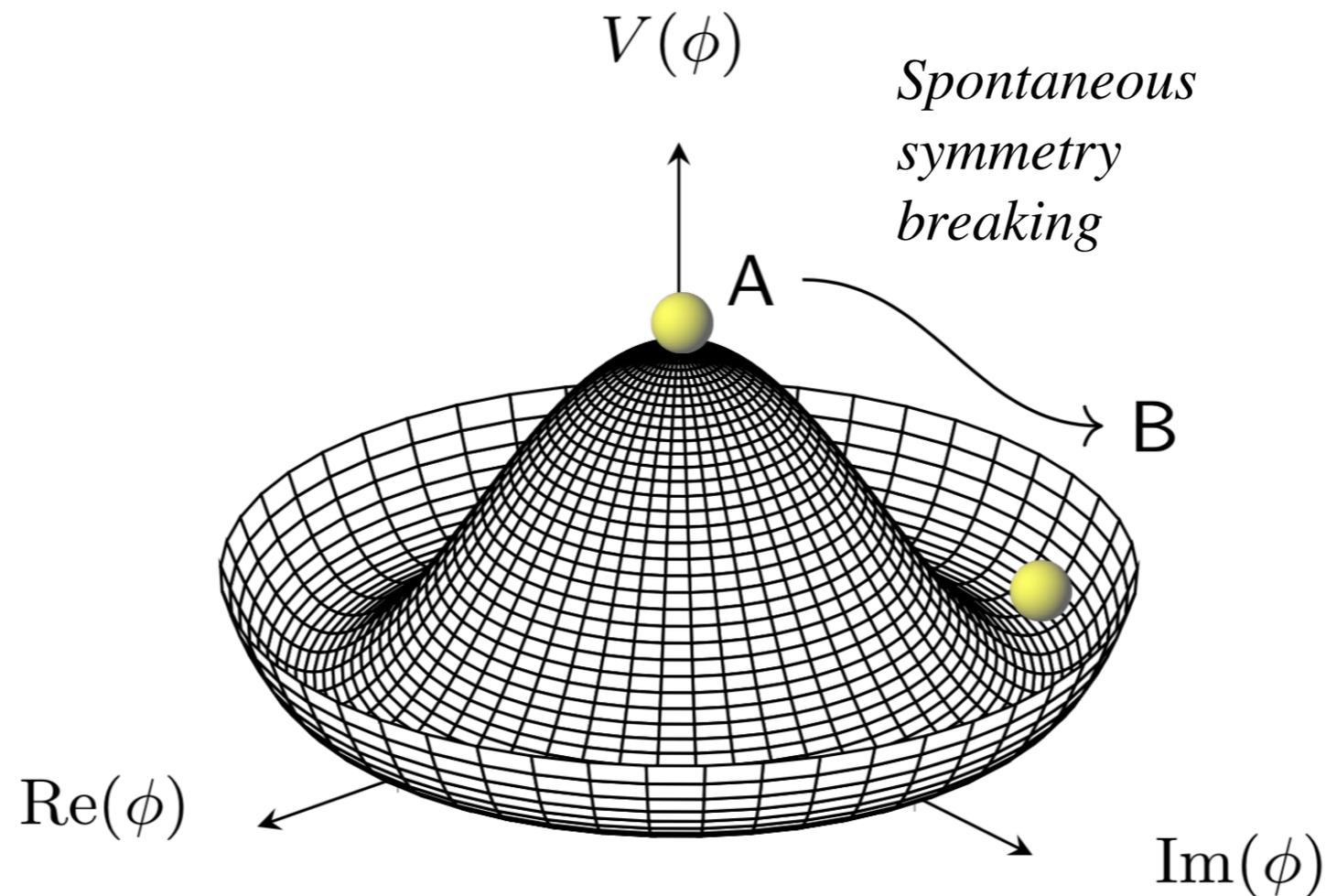


SIMON FRASER
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ENGAGING THE WORLD

The Higgs-boson story (in a tiny nutshell)

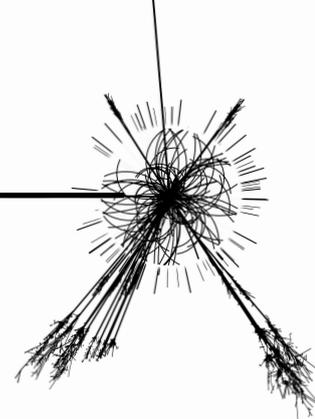


- ▶ Problem: No explanation of boson masses in theory
- ▶ 1964: Brout-Englert-**Higgs** mechanism with prediction of new particle

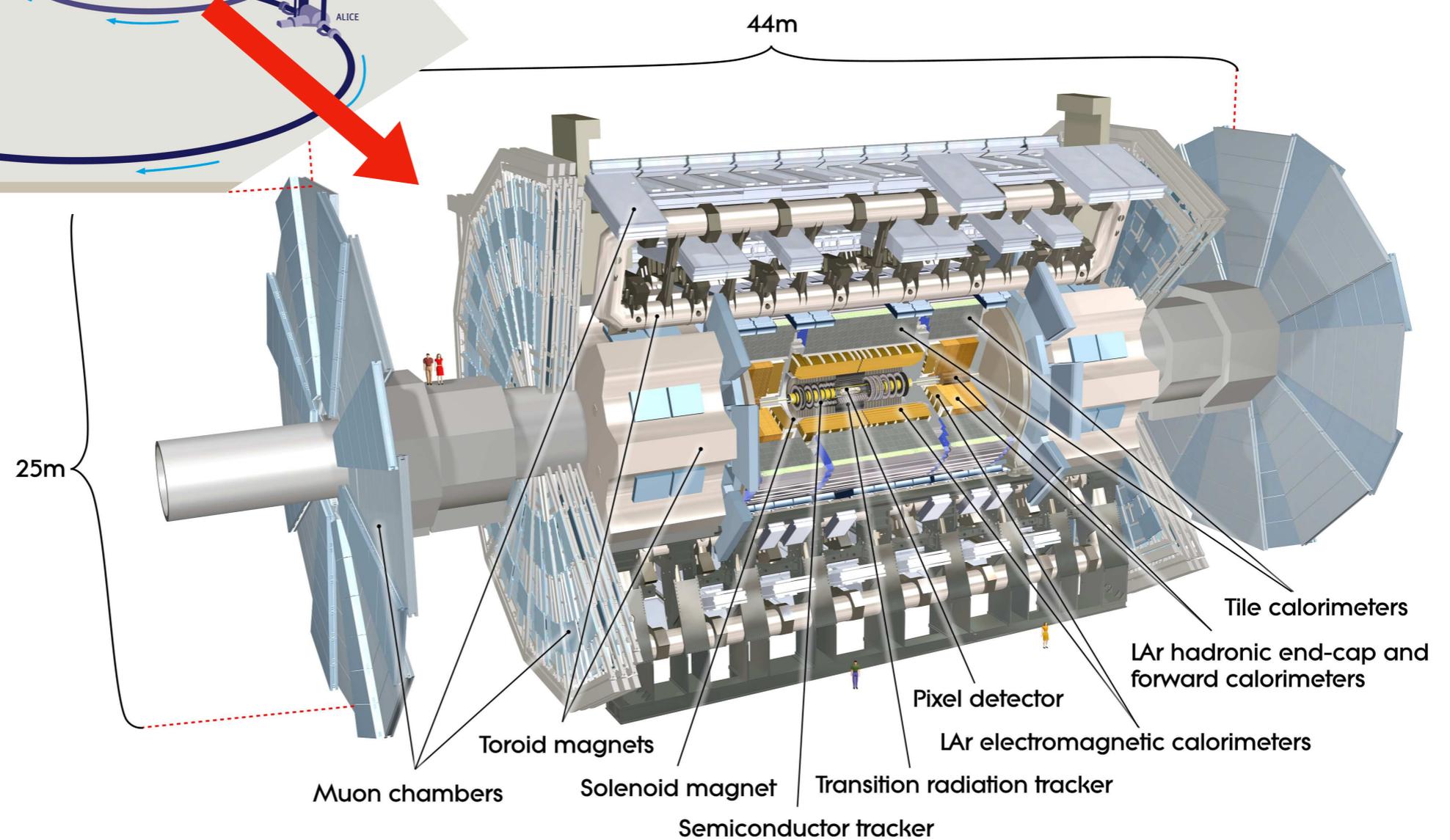
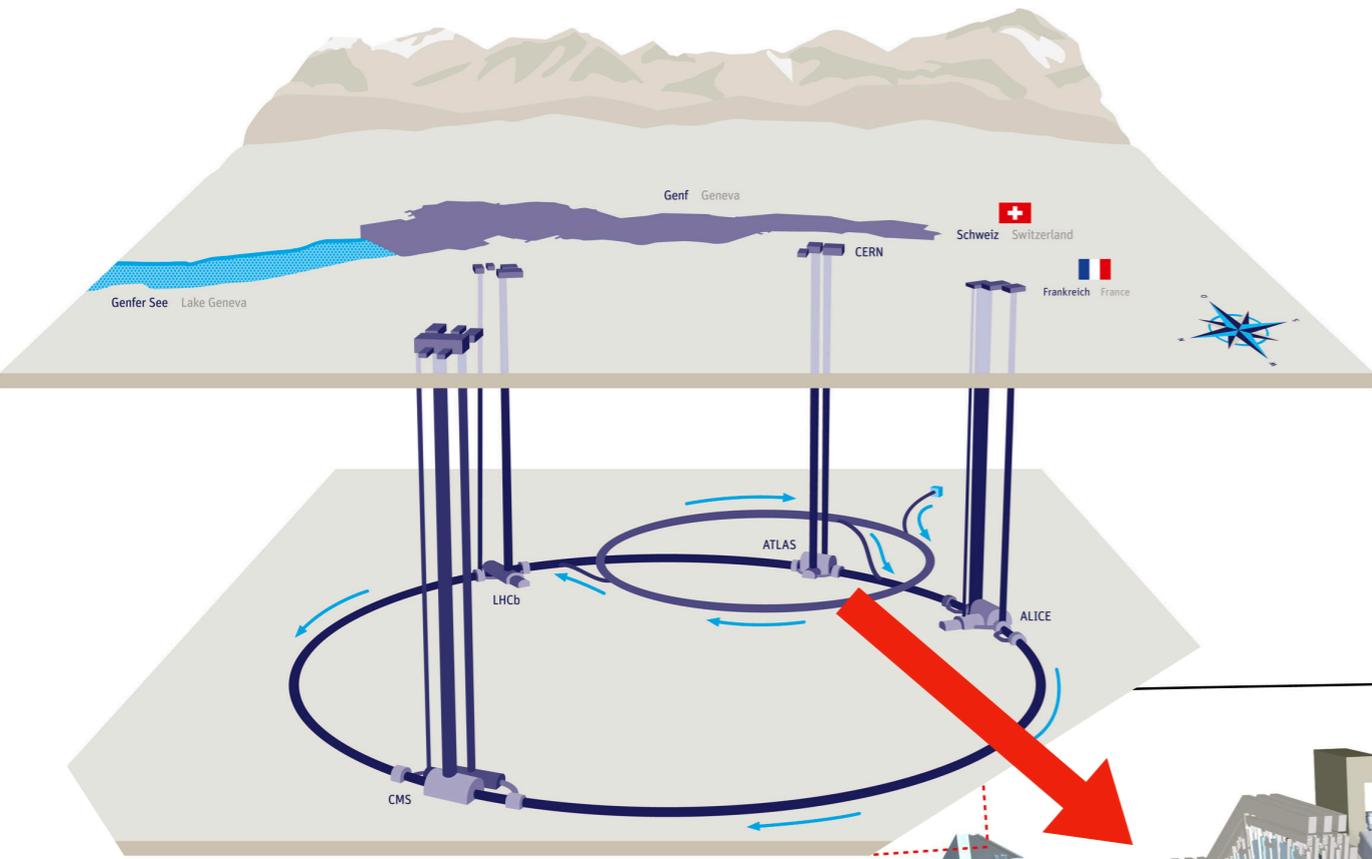


- ▶ 2012: Experimental **discovery** of Higgs boson by ATLAS and CMS collaborations at the LHC

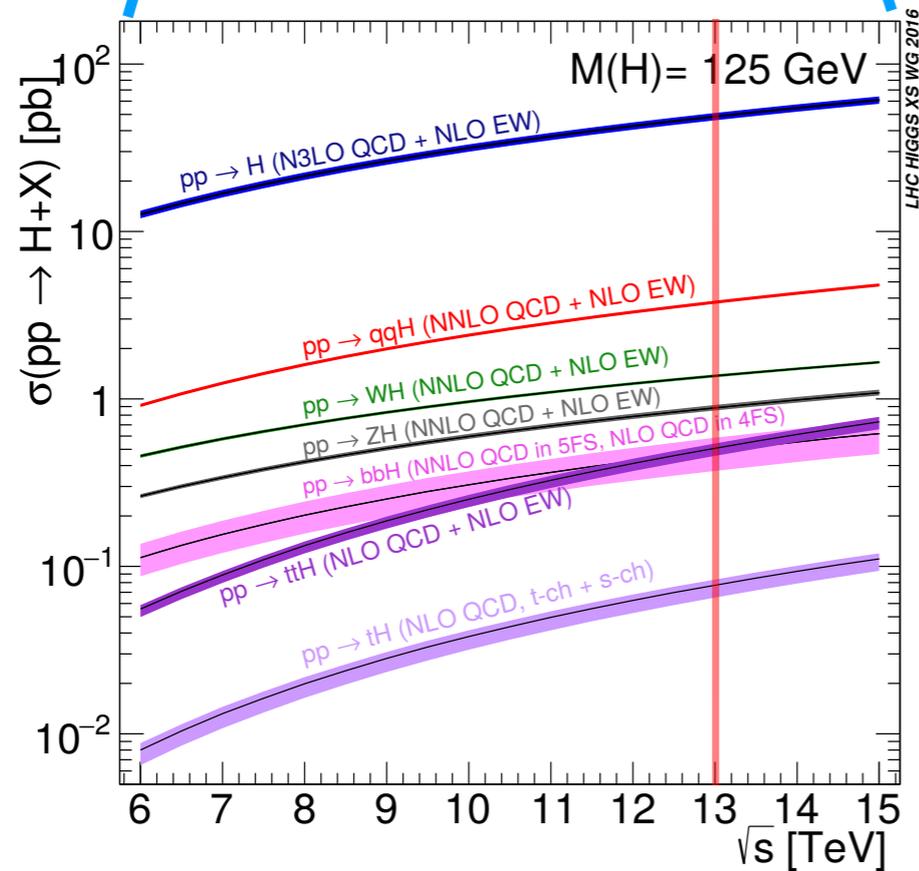
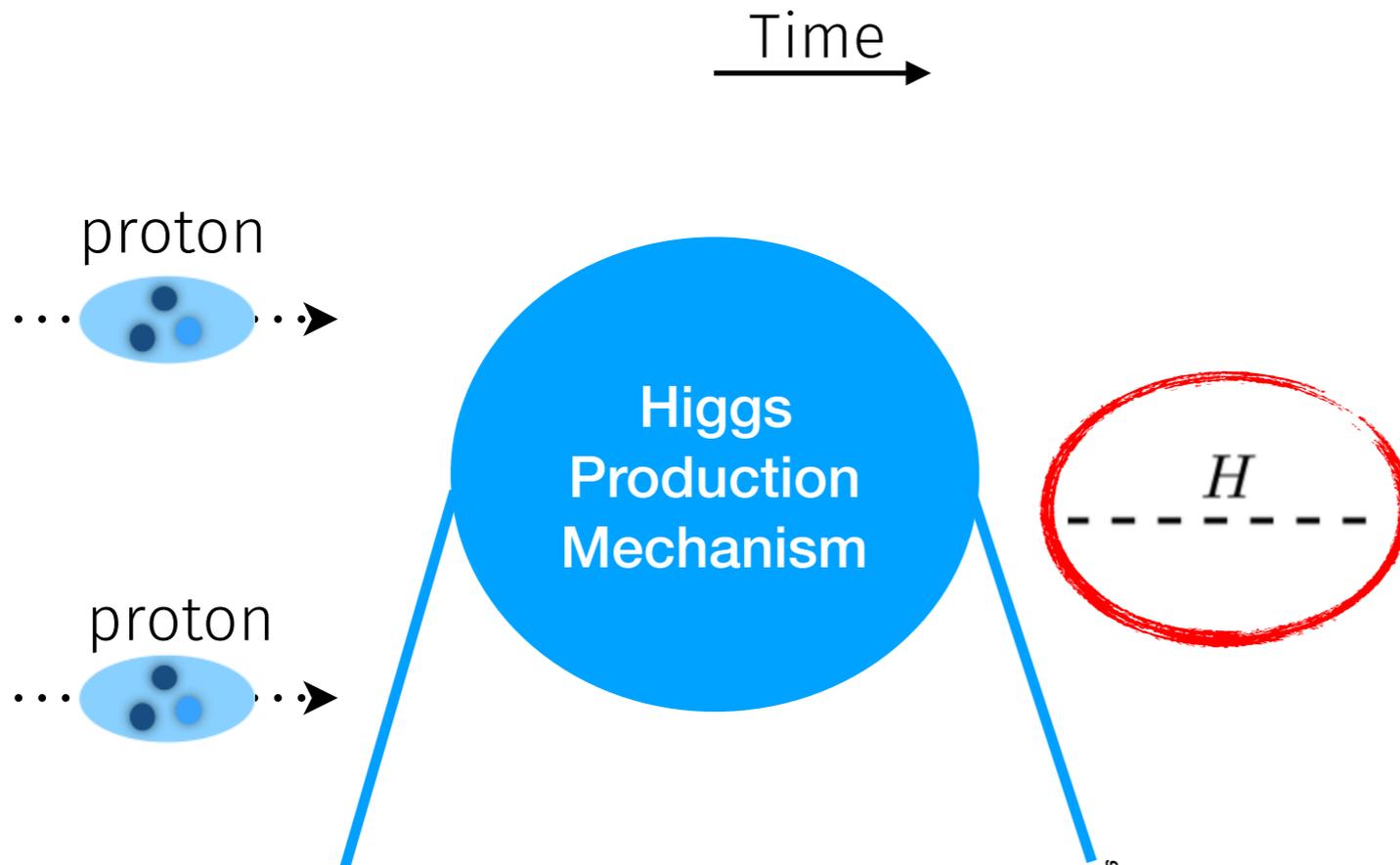
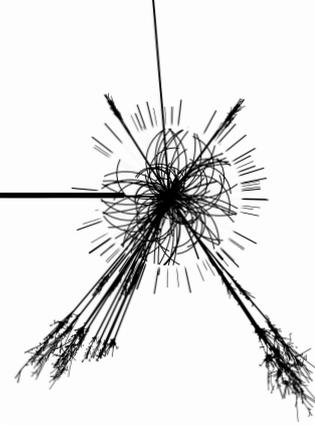
The LHC and the ATLAS Experiment



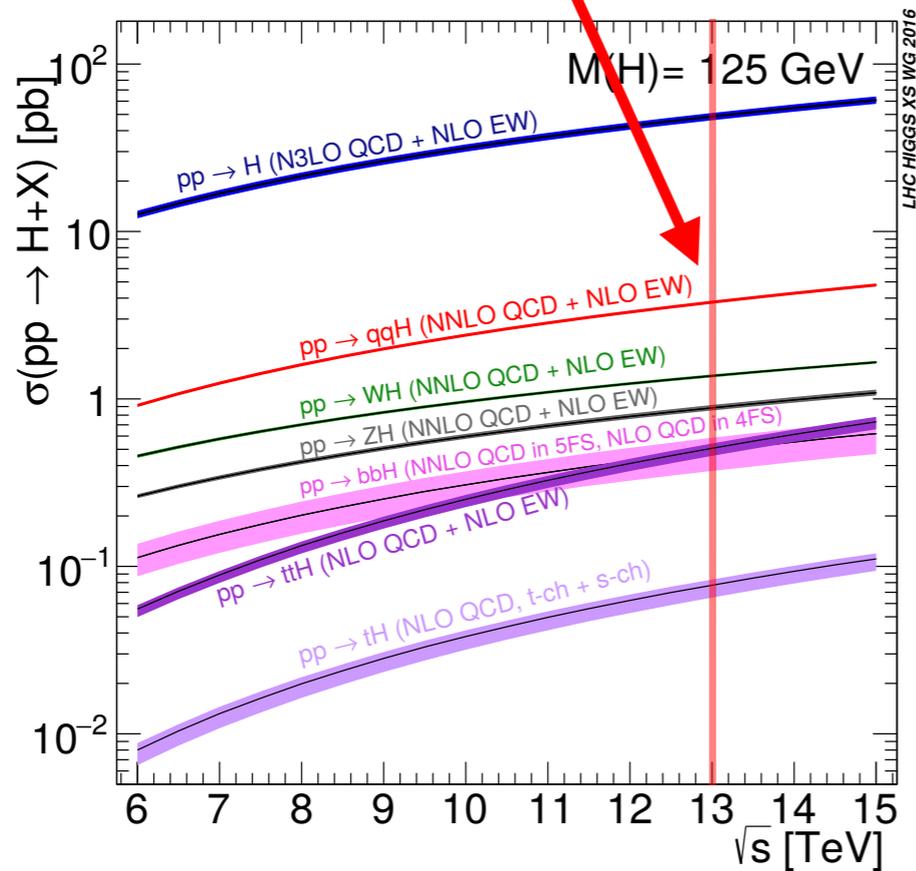
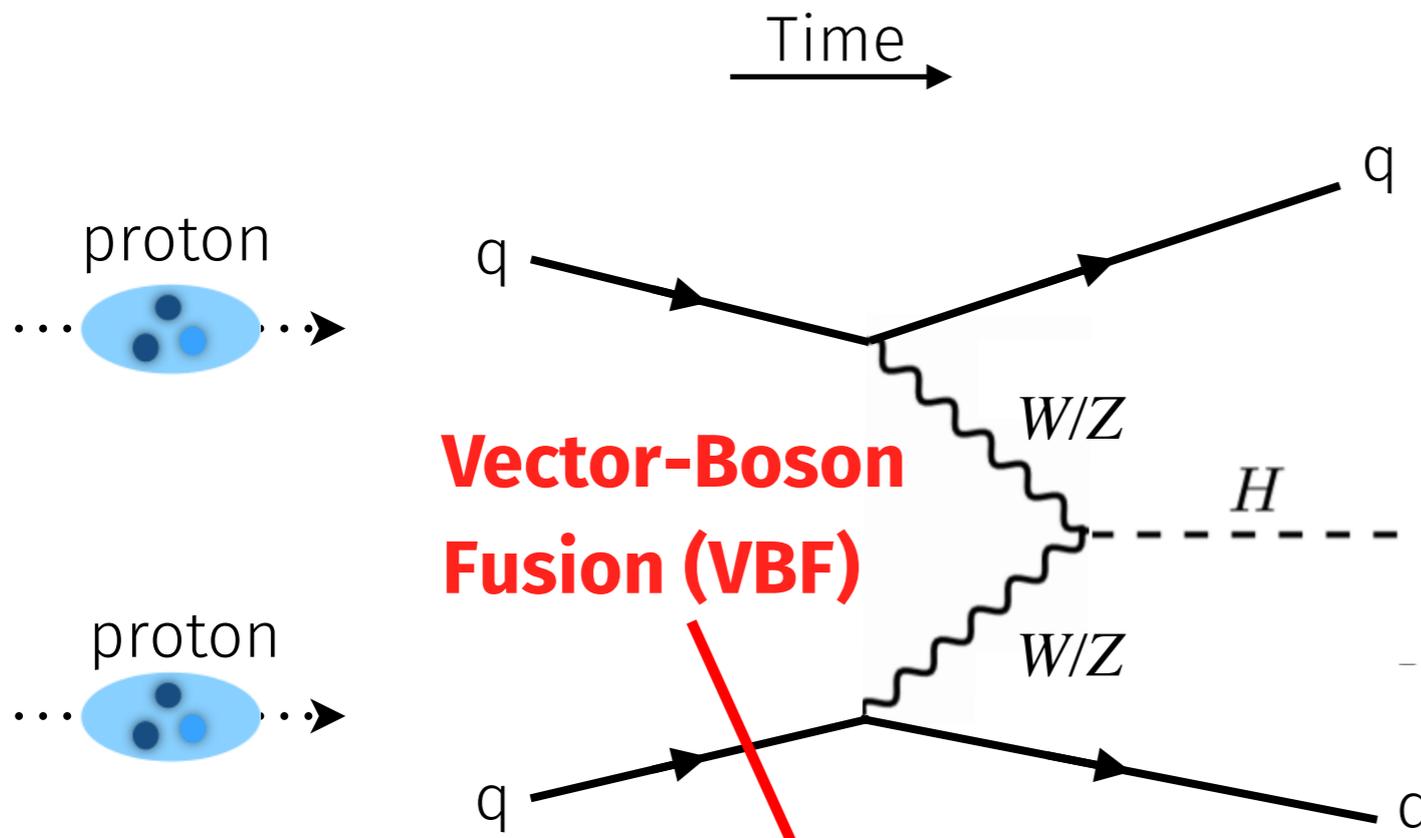
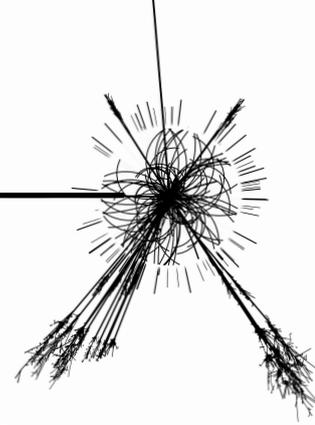
- ▶ LHC provides high-energy proton-proton collisions
- ▶ Detected by different detectors e.g. the ATLAS experiment



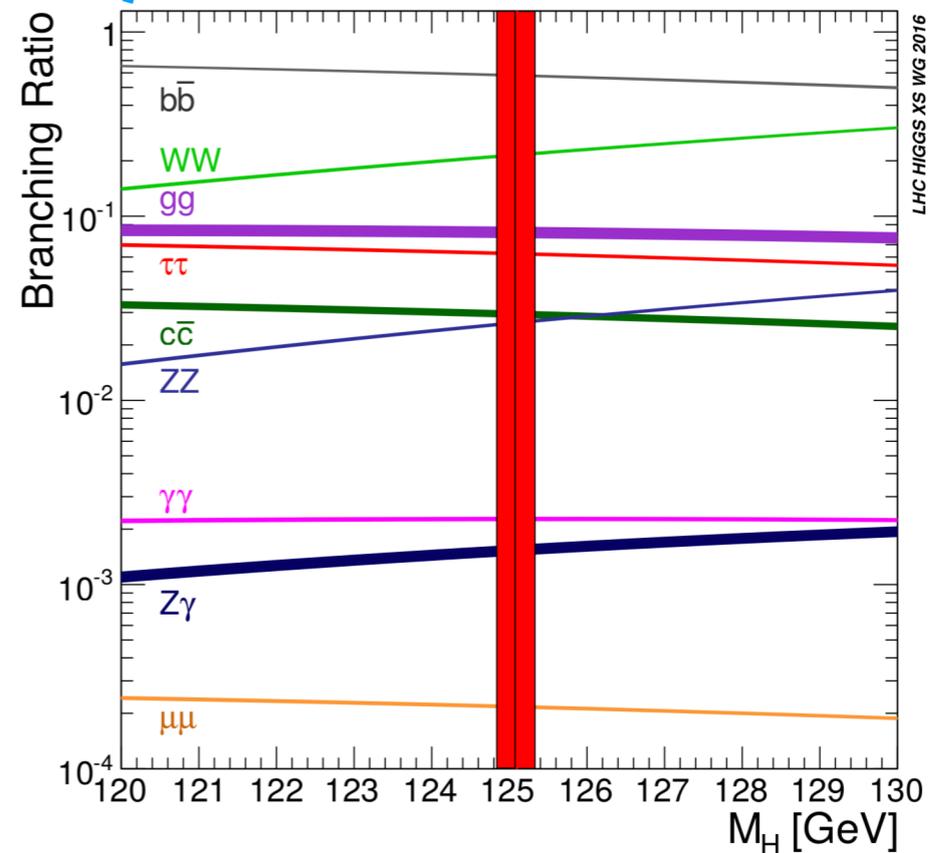
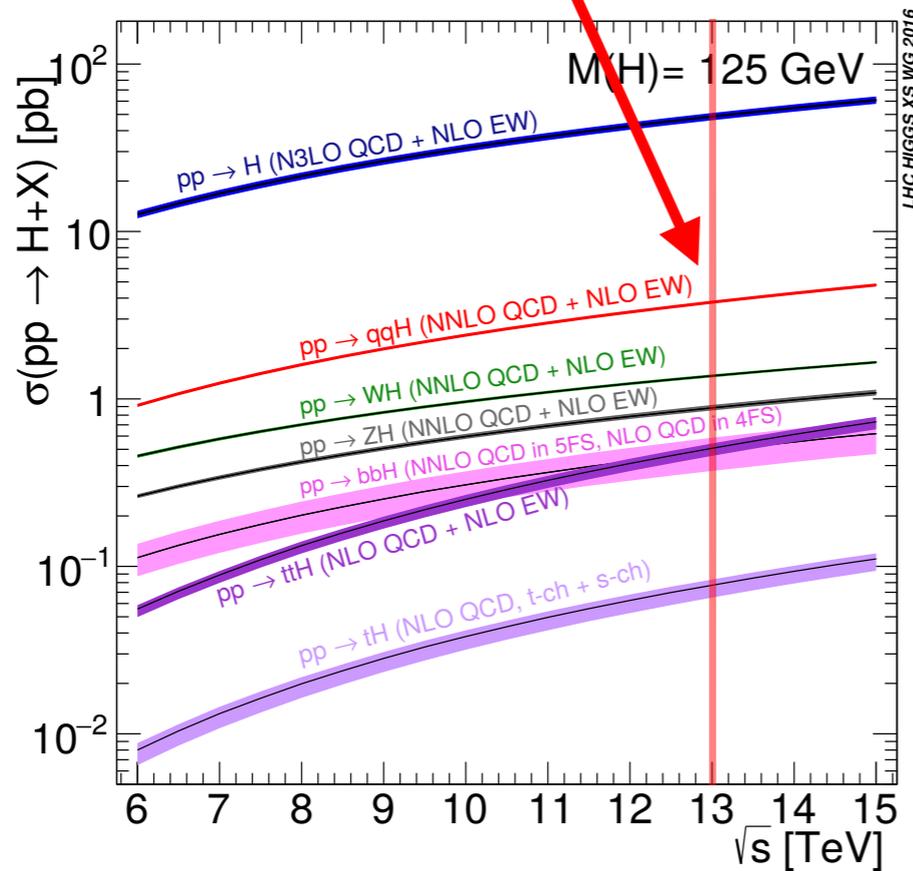
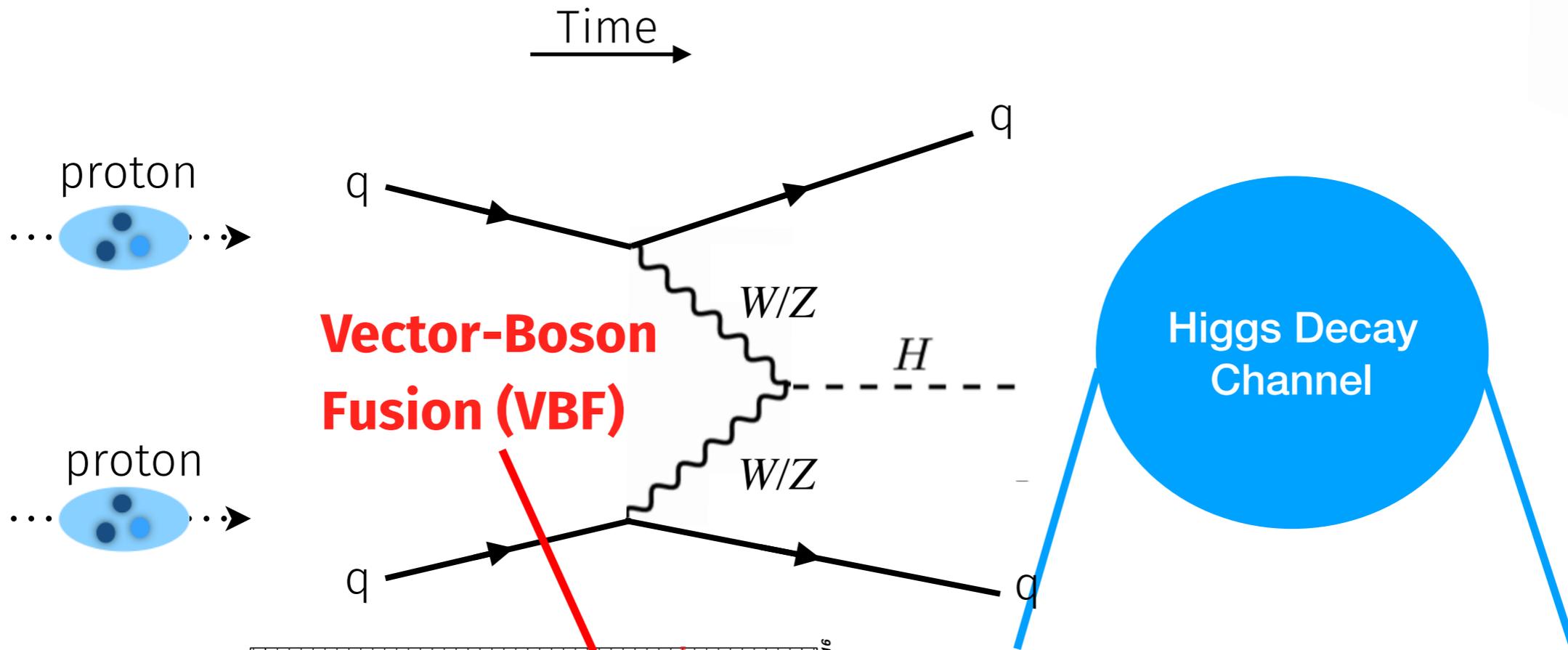
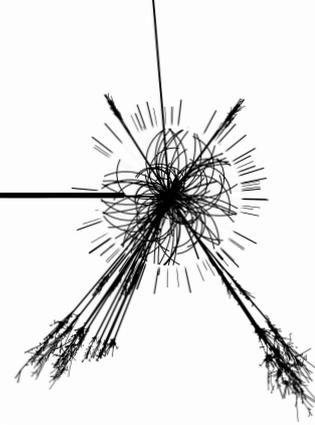
The VBF Higgs to WW Process



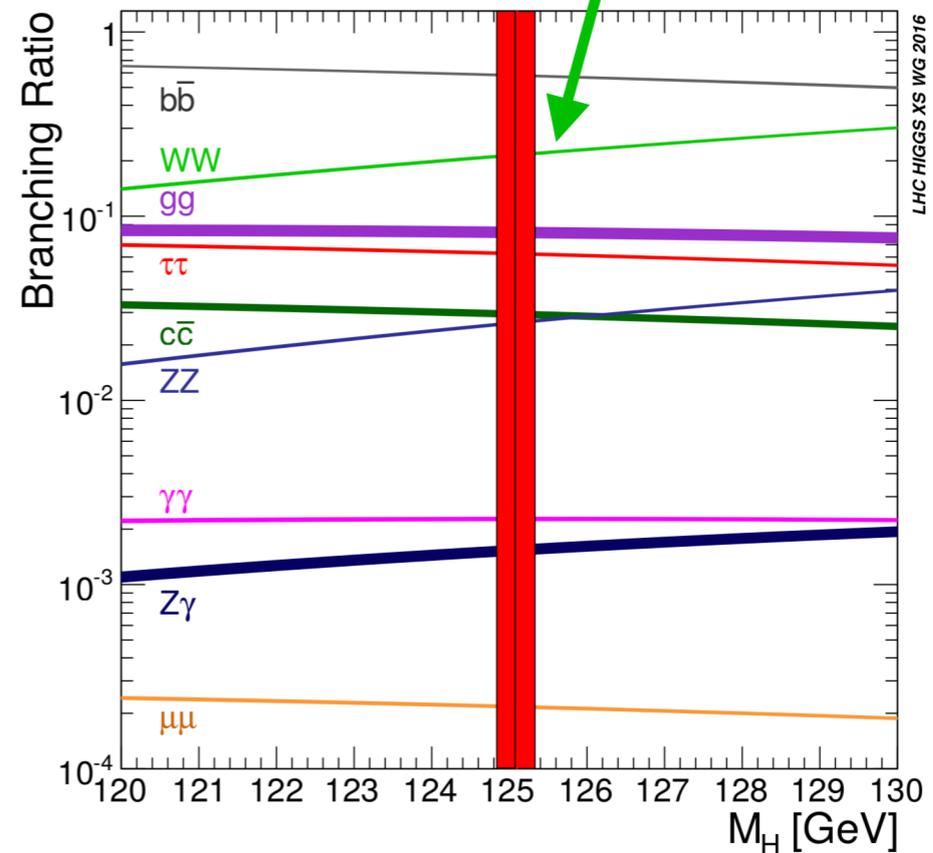
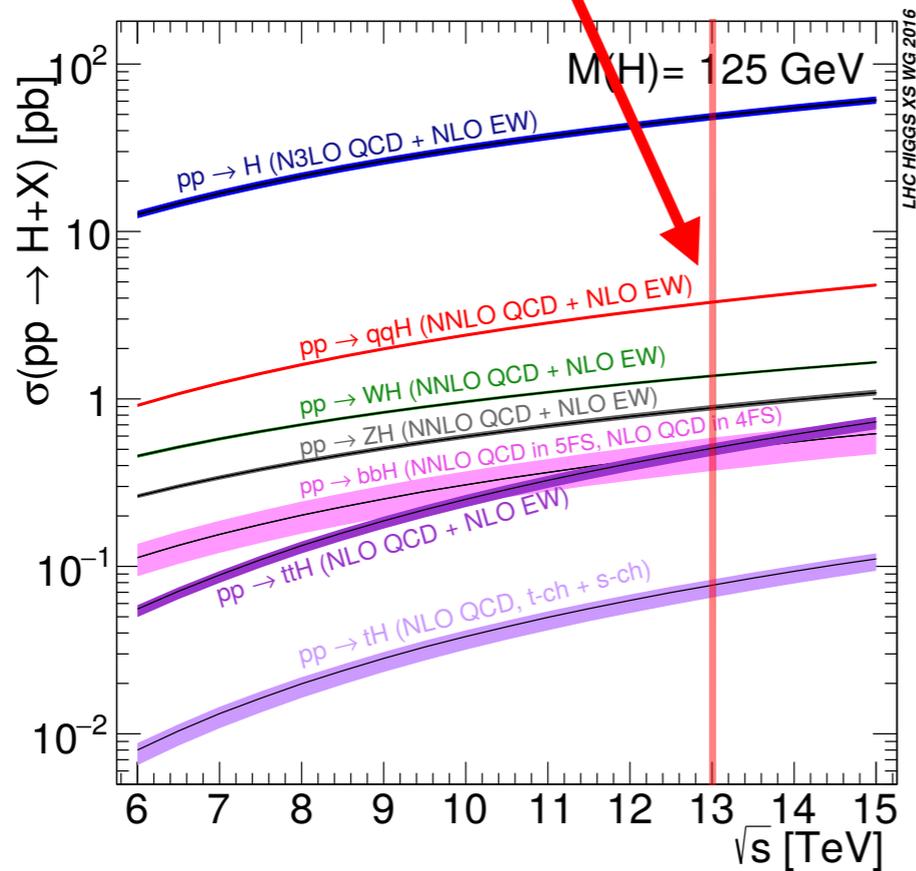
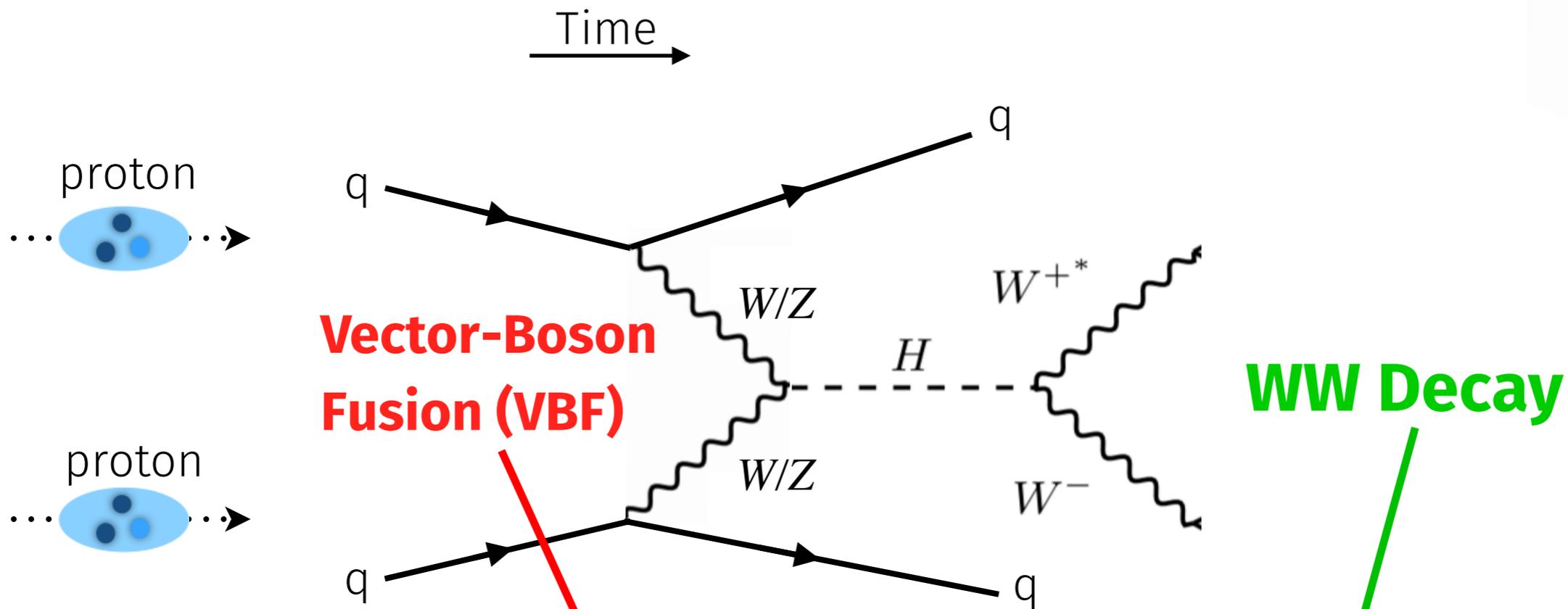
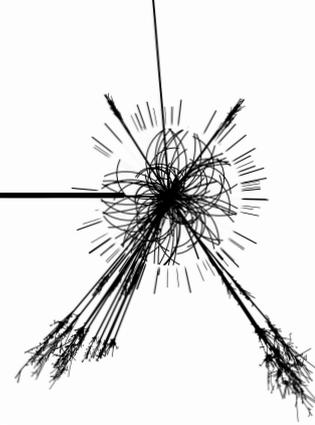
The VBF Higgs to WW Process



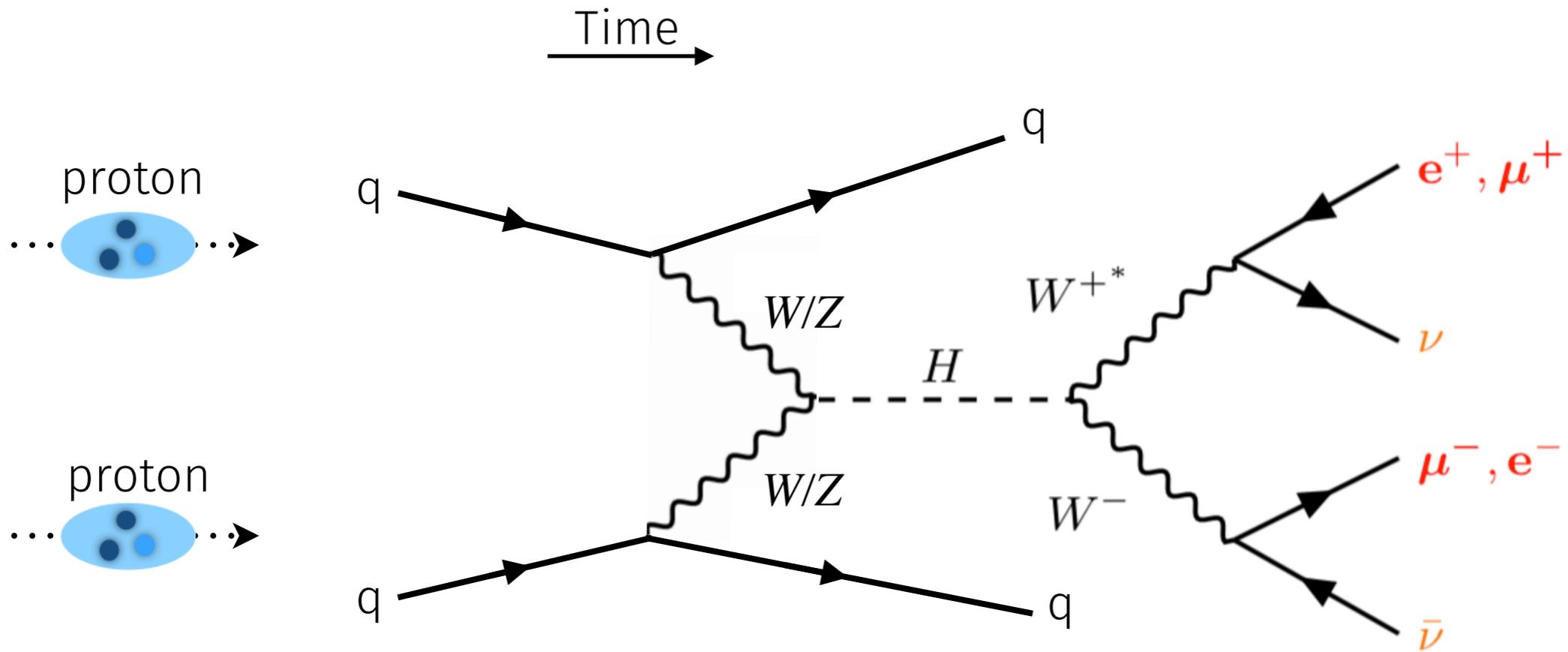
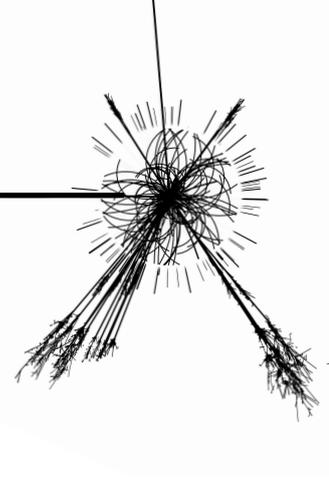
The VBF Higgs to WW Process



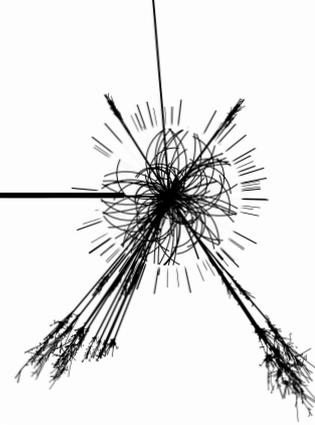
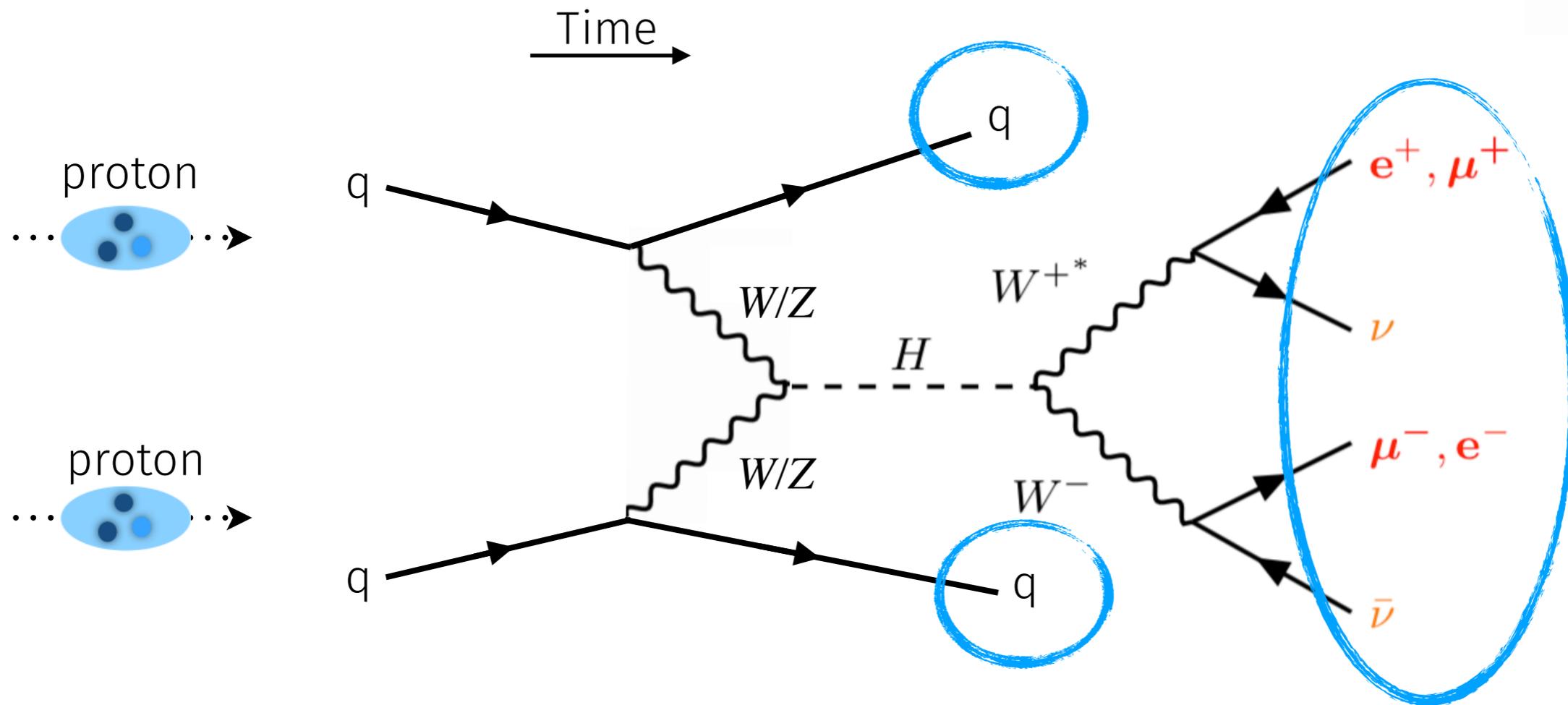
The VBF Higgs to WW Process



The VBF Higgs to WW Process



The VBF Higgs to WW Process

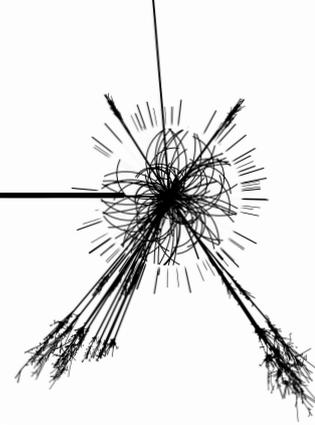


Analyzed **final state** has...

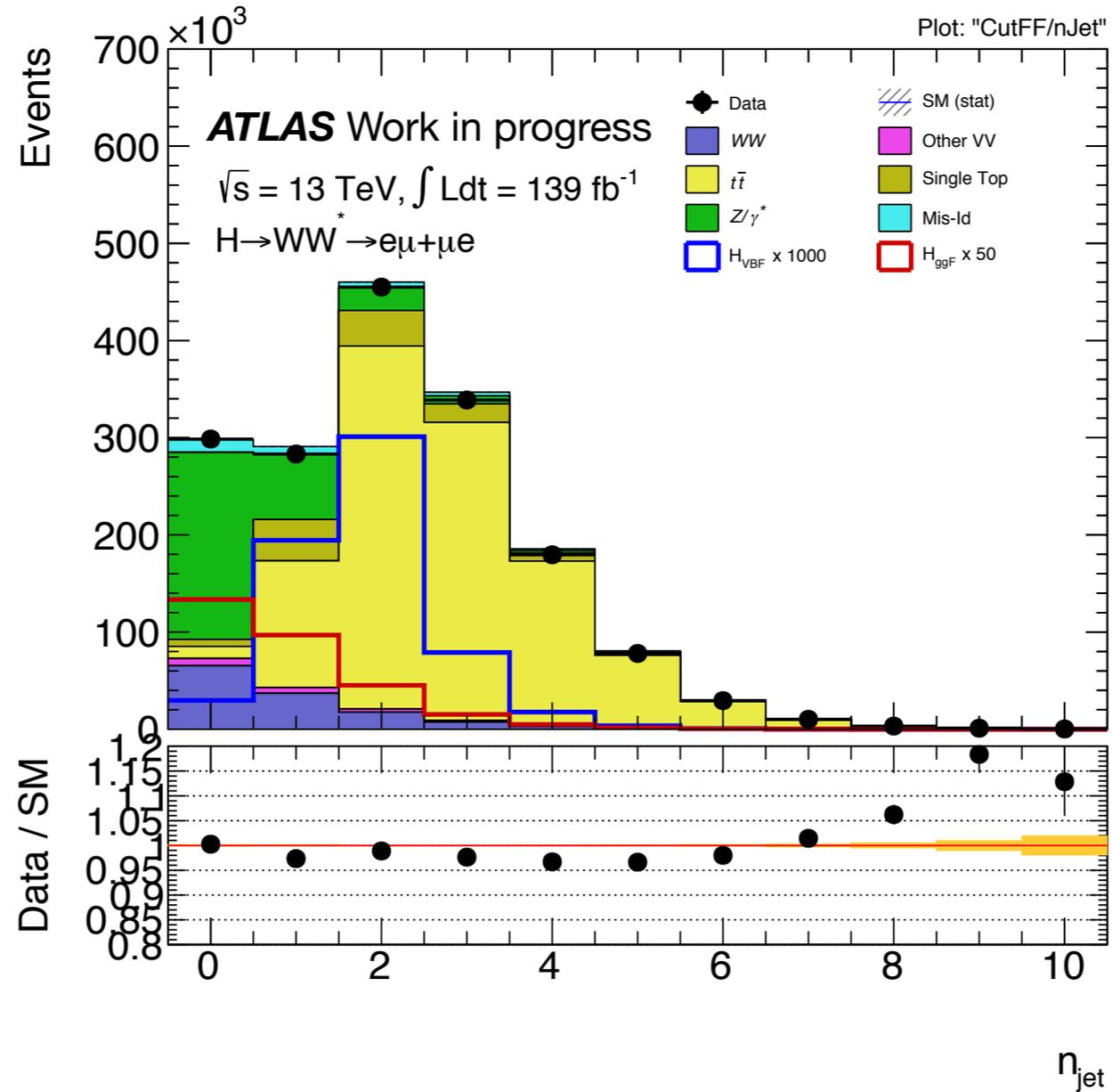
- ▶ 2 jets (from quarks) in the forward region of the detector
- ▶ Two neutrinos
- ▶ Two isolated, oppositely charged, *different flavour leptons*

Drop same flavour channel due to overwhelming background

Contributing Backgrounds

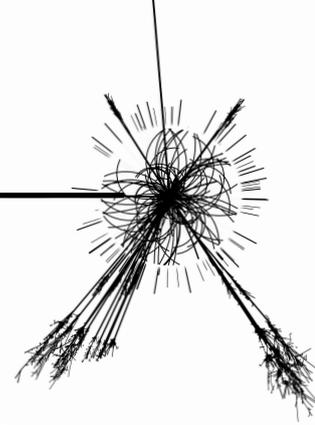


- ▶ Several other types of processes that mimic the Higgs signature (a nuisance!)



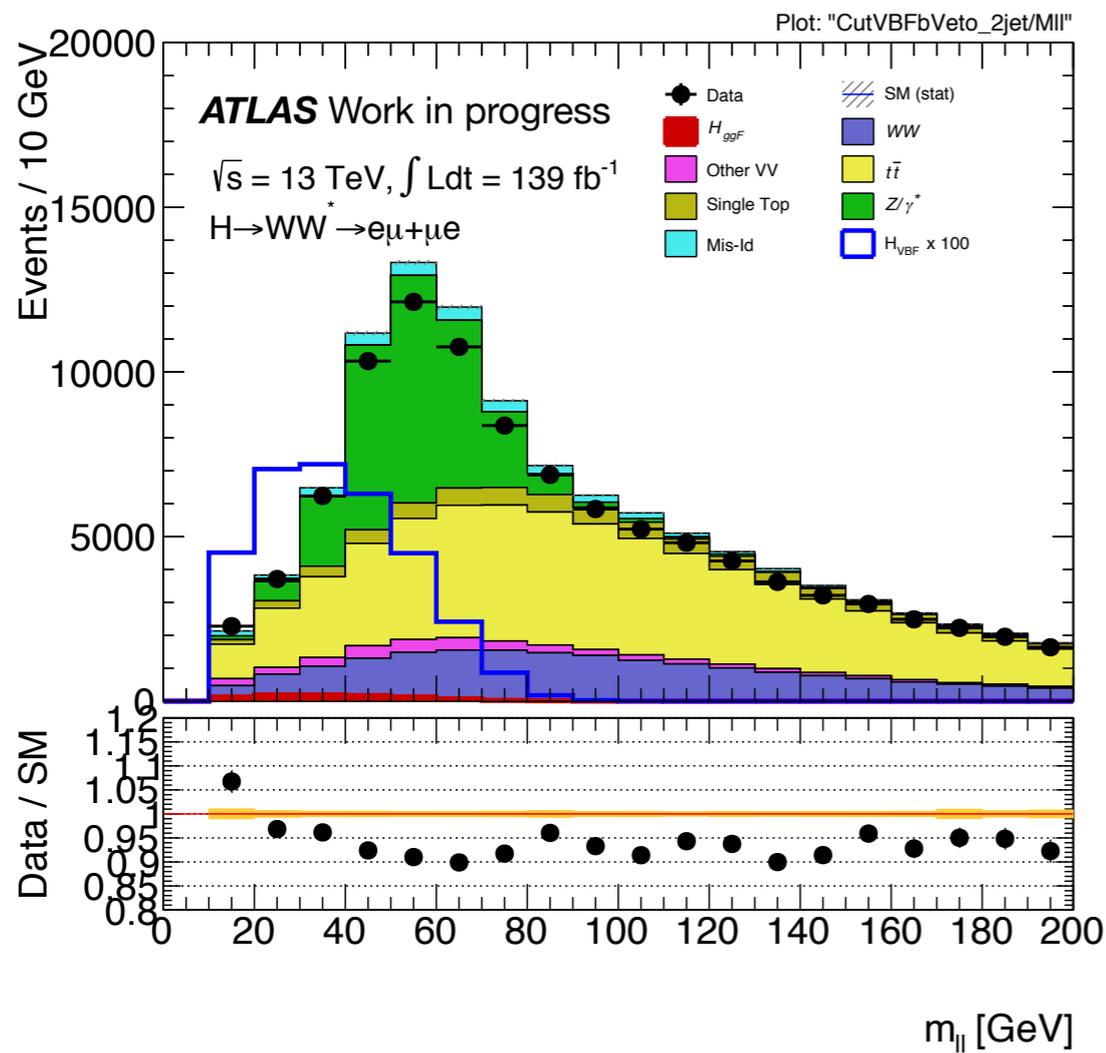
- ▶ Need to select a region (**kinematic/topological** selections) that is as pure in Higgs events as possible
- ▶ Suitable task for a **neural network** to classify events

Train a Deep Neural Network for Event Classification (1)

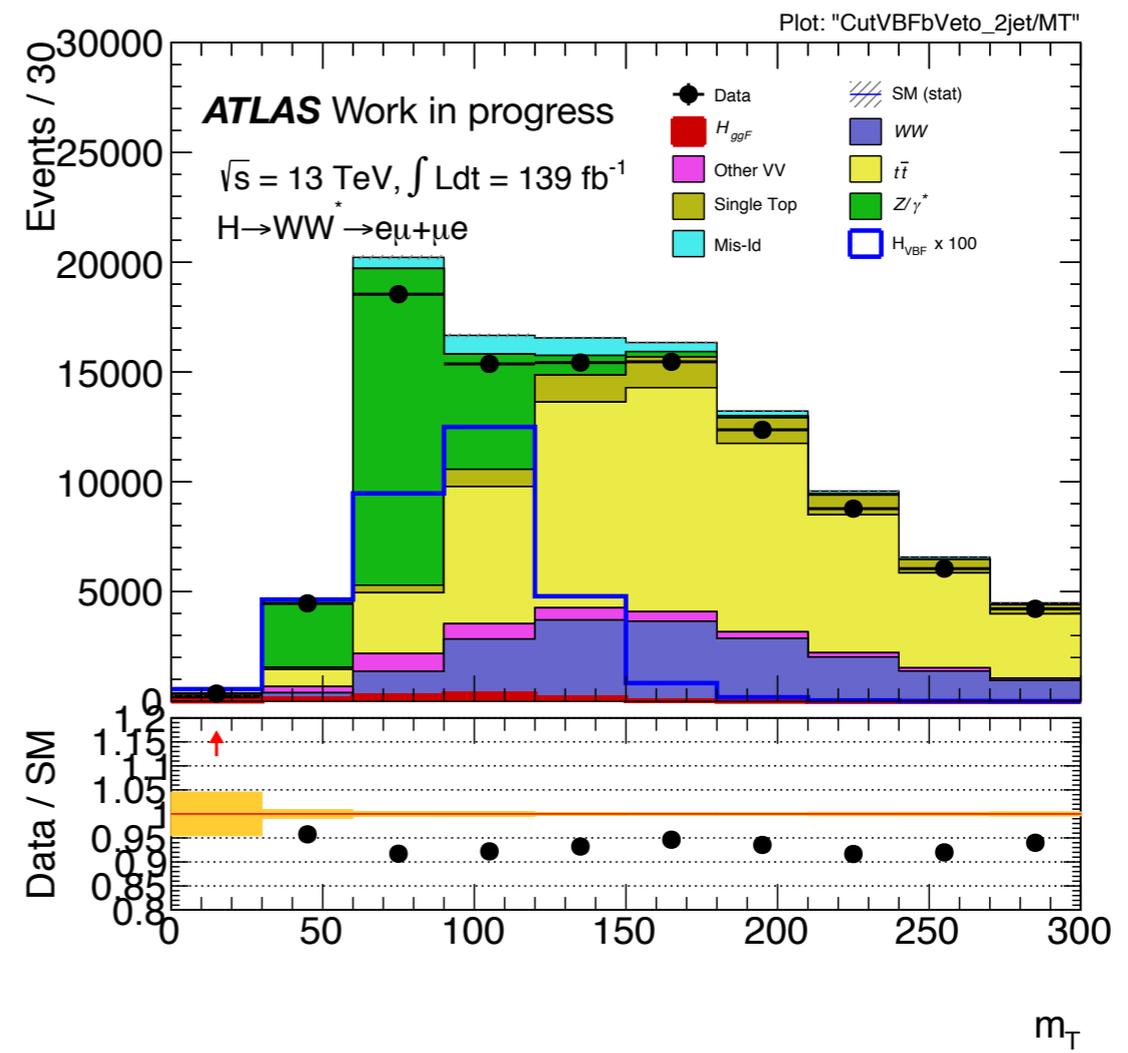


► **15 input variables** used in the training

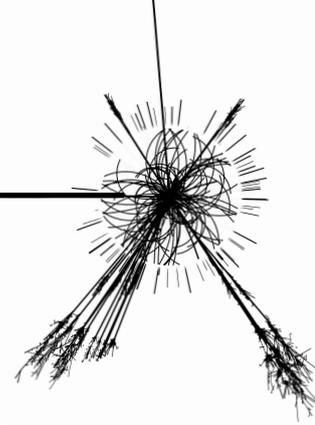
► Mass of the lepton system



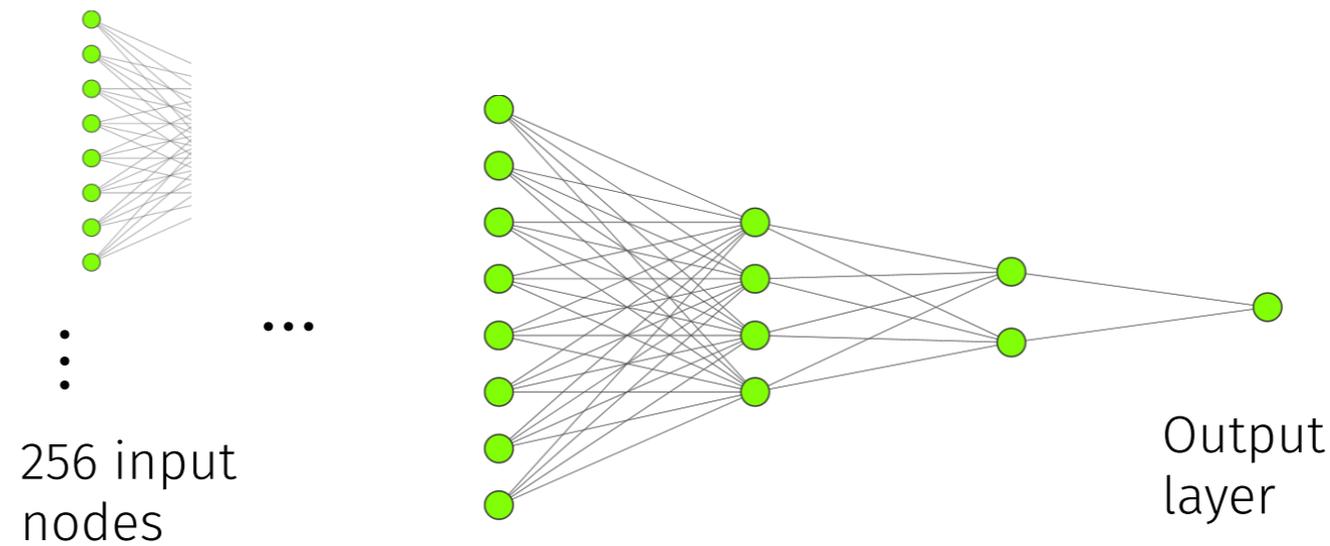
► Transverse mass of the whole system



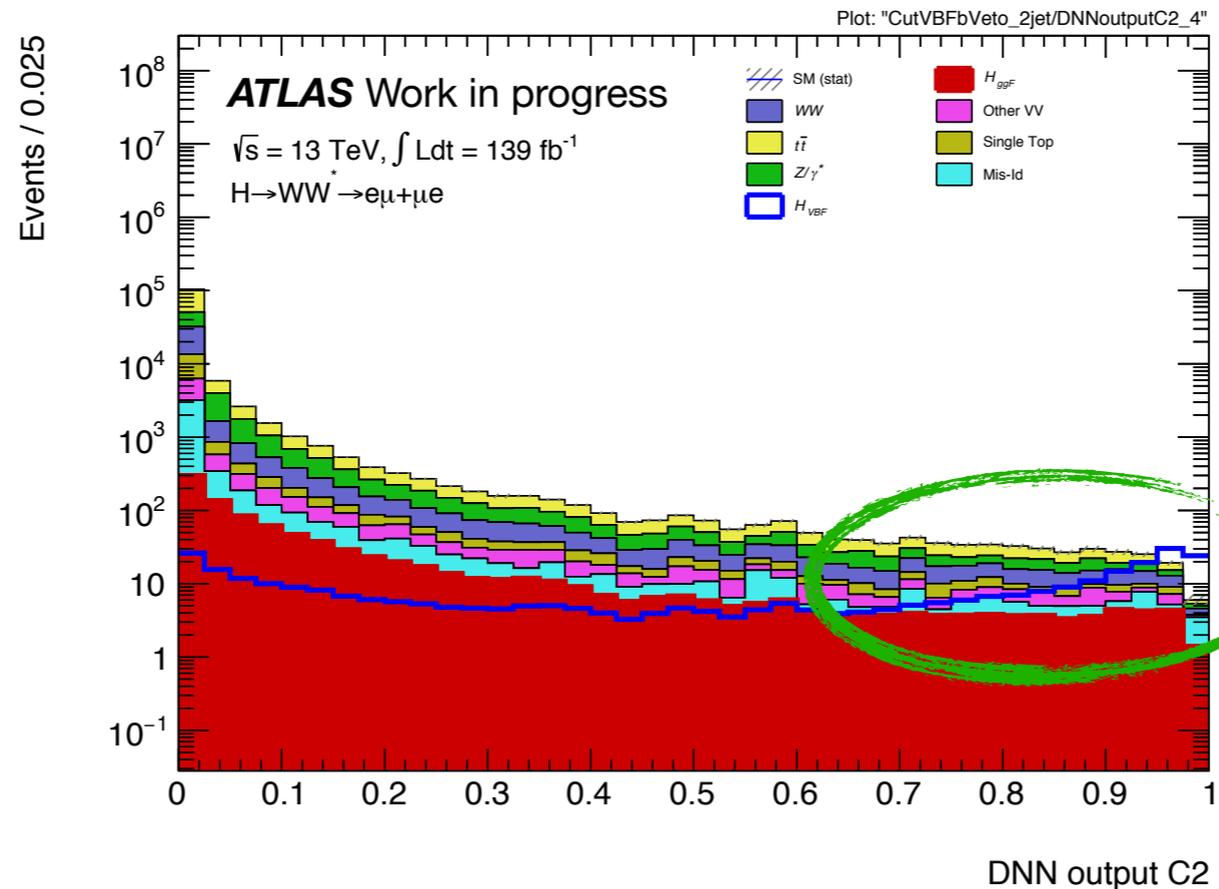
Train a Deep Neural Network for Event Classification (2)



► New “deep” network architecture (8 layers)



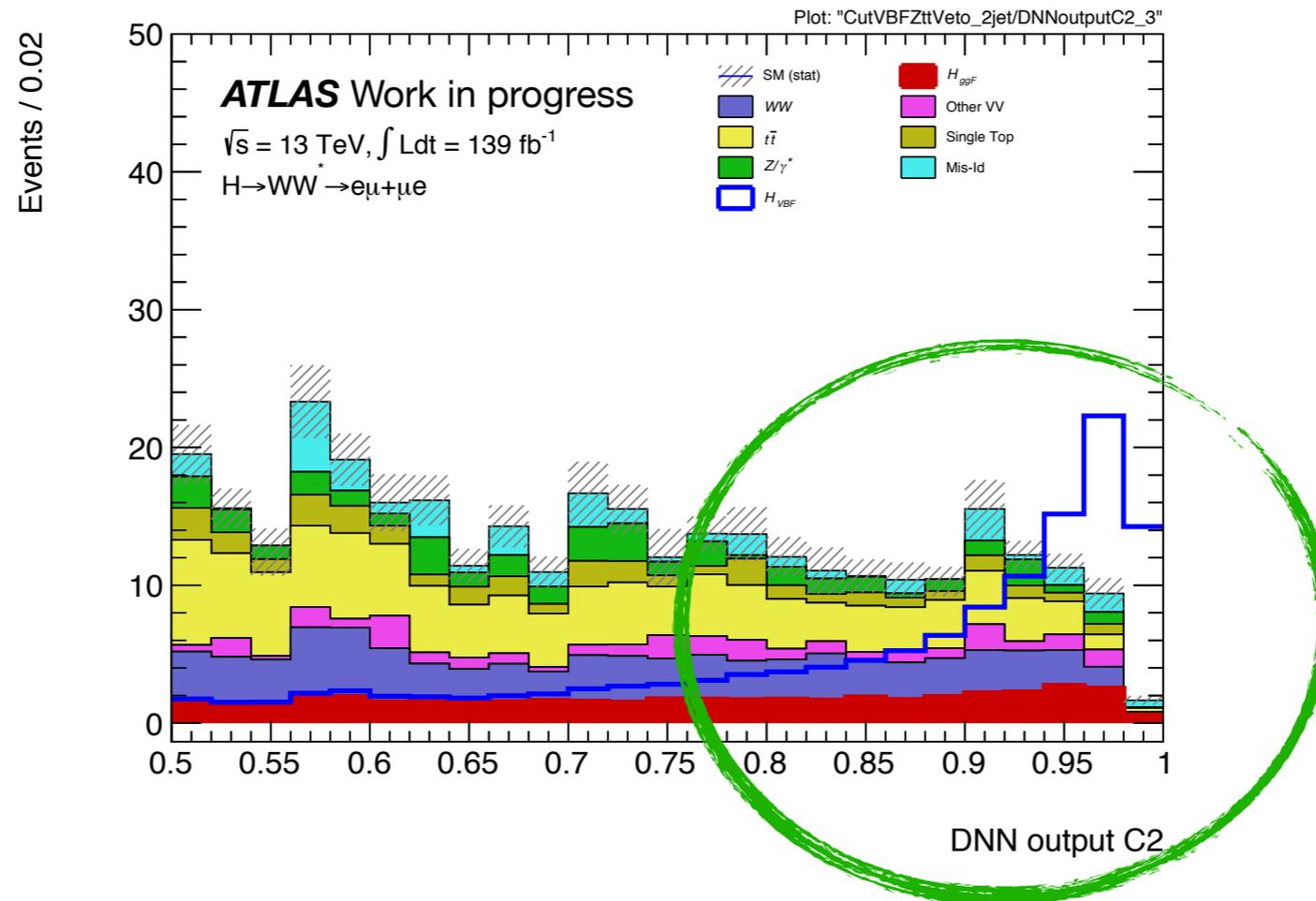
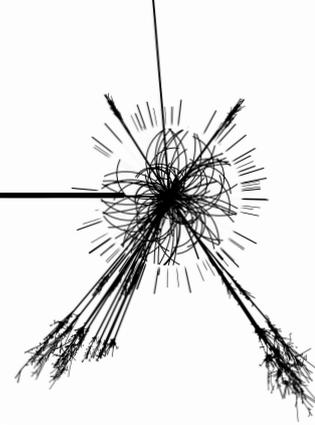
► DNN output distribution



Signal sensitive region

... let's zoom in

Output of Neural Network



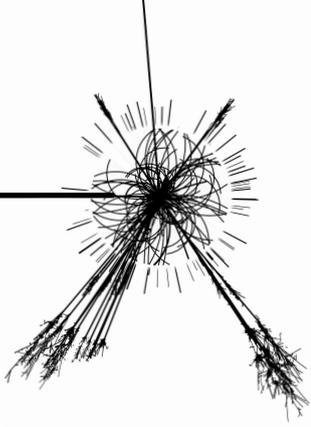
- ▶ How to estimate the performance of the DNN classifier?
- ▶ Simple estimators for the expected discovery significance can be used:

$$Z = \frac{s}{\sqrt{b}}$$

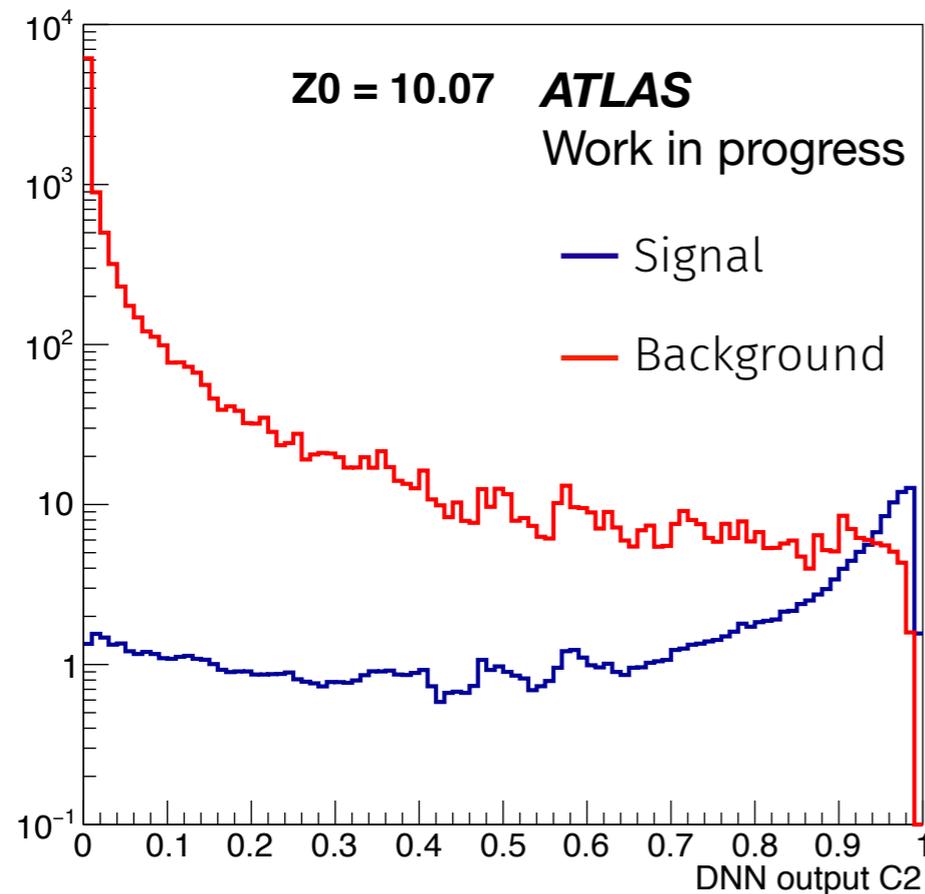
$$Z = \begin{cases} +\sqrt{2 \left(n \ln \left[\frac{n(b+\sigma^2)}{b^2+n\sigma^2} \right] - \frac{b^2}{\sigma^2} \ln \left[1 + \frac{\sigma^2(n-b)}{b(b+\sigma^2)} \right] \right)} & \text{if } n \geq b \\ -\sqrt{2 \left(n \ln \left[\frac{n(b+\sigma^2)}{b^2+n\sigma^2} \right] - \frac{b^2}{\sigma^2} \ln \left[1 + \frac{\sigma^2(n-b)}{b(b+\sigma^2)} \right] \right)} & \text{if } n < b. \end{cases}$$

- ▶ Bins are statistically independent so add them in quadrature: $Z = \sum_i z_{\text{bin } i}^2$

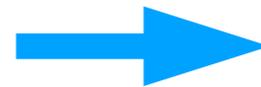
Binning Optimization Strategies



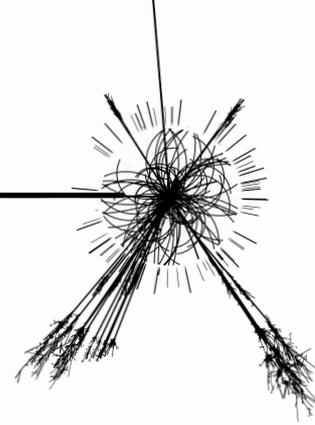
- ▶ For a stable statistical analysis, we need to reduce the number of bins
- ▶ Goal is to **exploit the shape** as well as possible
- ▶ In general: choose **as many bins as possible** with given statistics and uncertainties



Rebinning

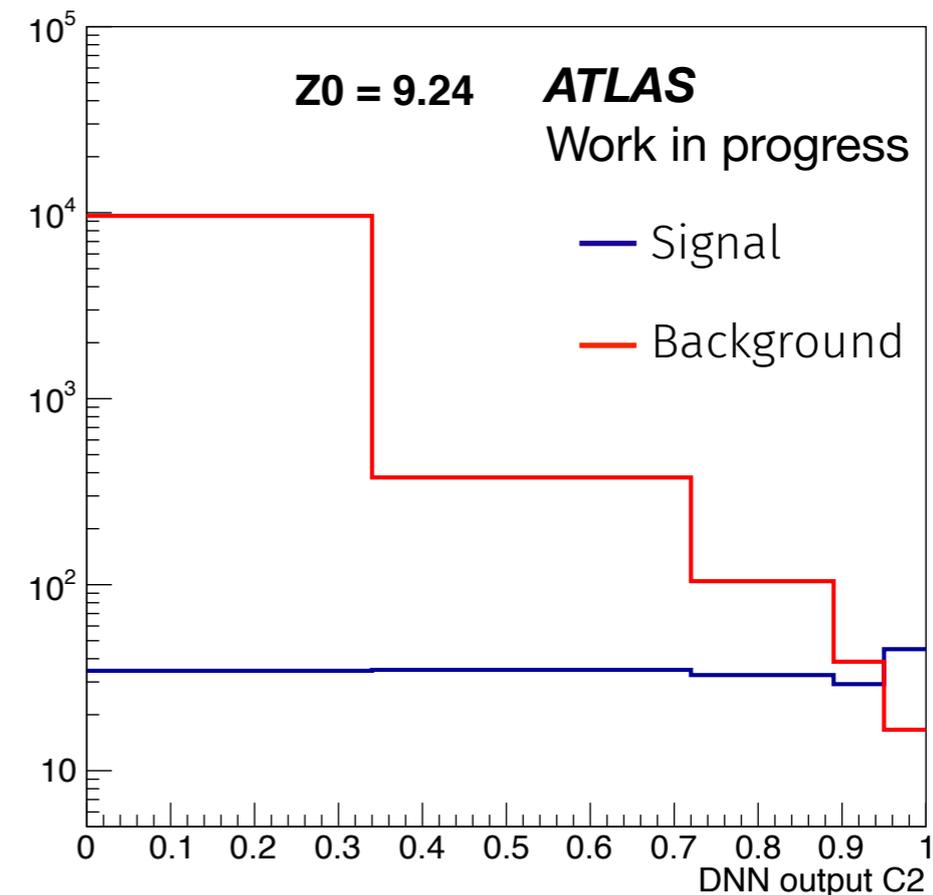
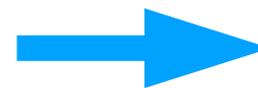
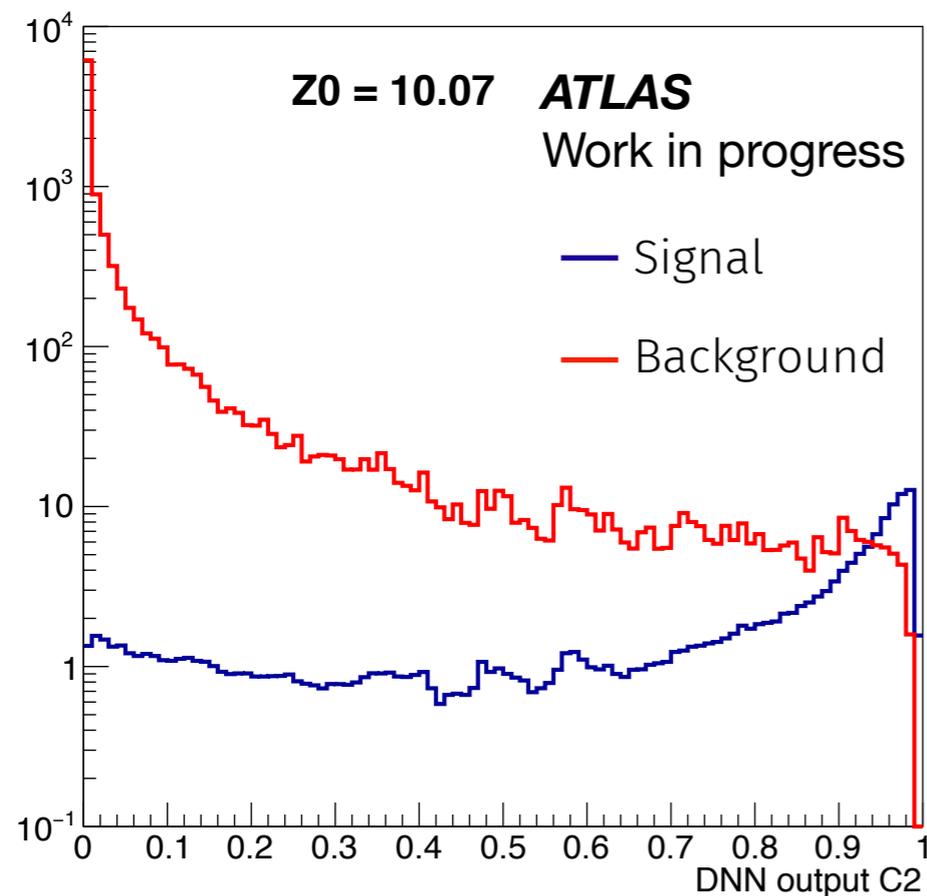


Binning Optimization Method 1

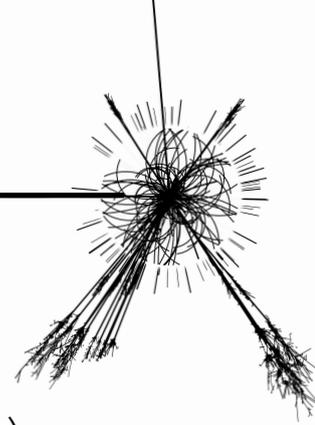


► Method 1: **Flat signal rebinning**

- Inputs: # of desired bins
- Strategy: choose bin boundaries such that in each bin a signal yield of **“total signal yield / # of bins”** is contained



Compare Different Methods



► Method 1

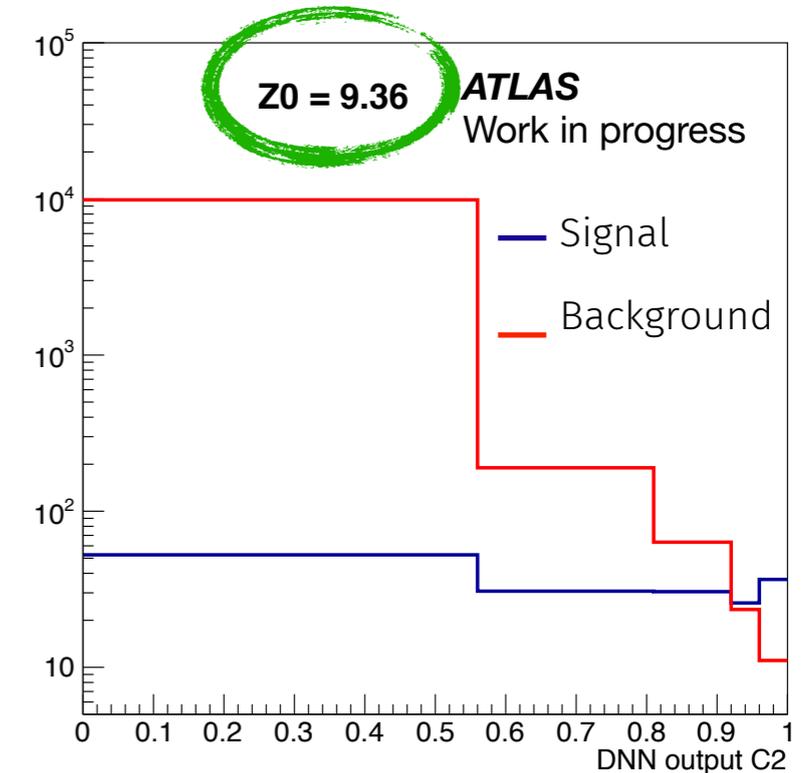
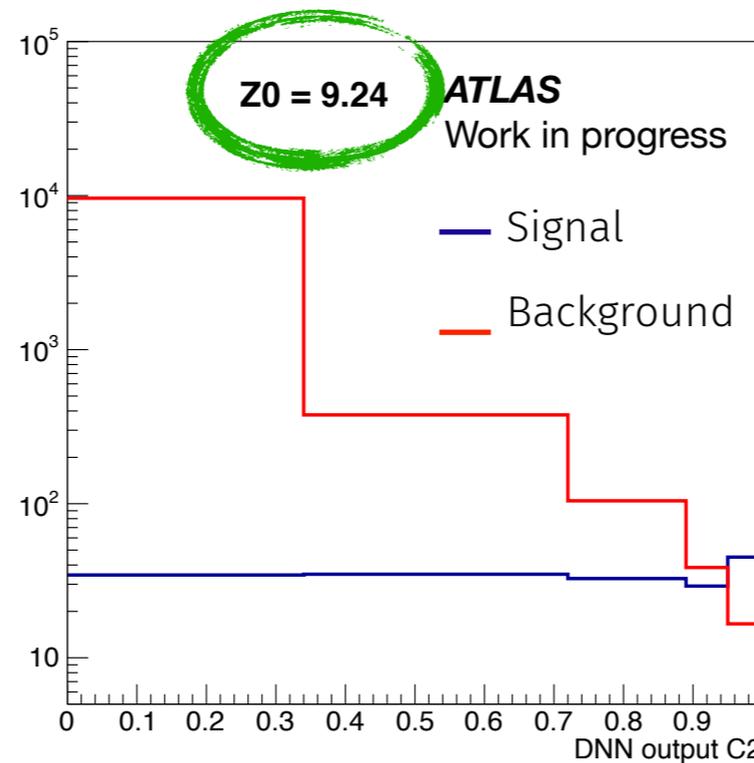
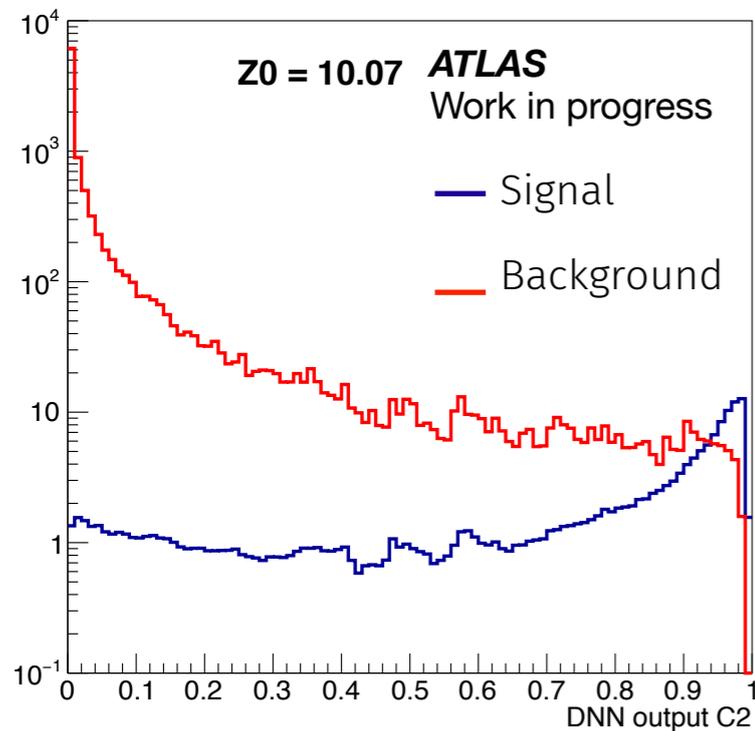
► Method 2

(explained only in backup)

Original

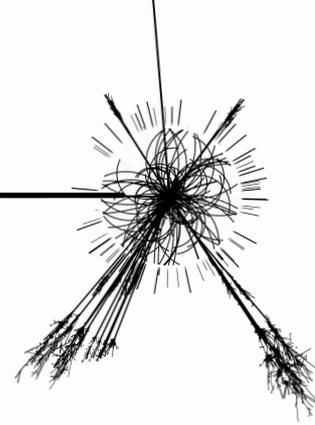
Flat signal rebinning

Significance optimization

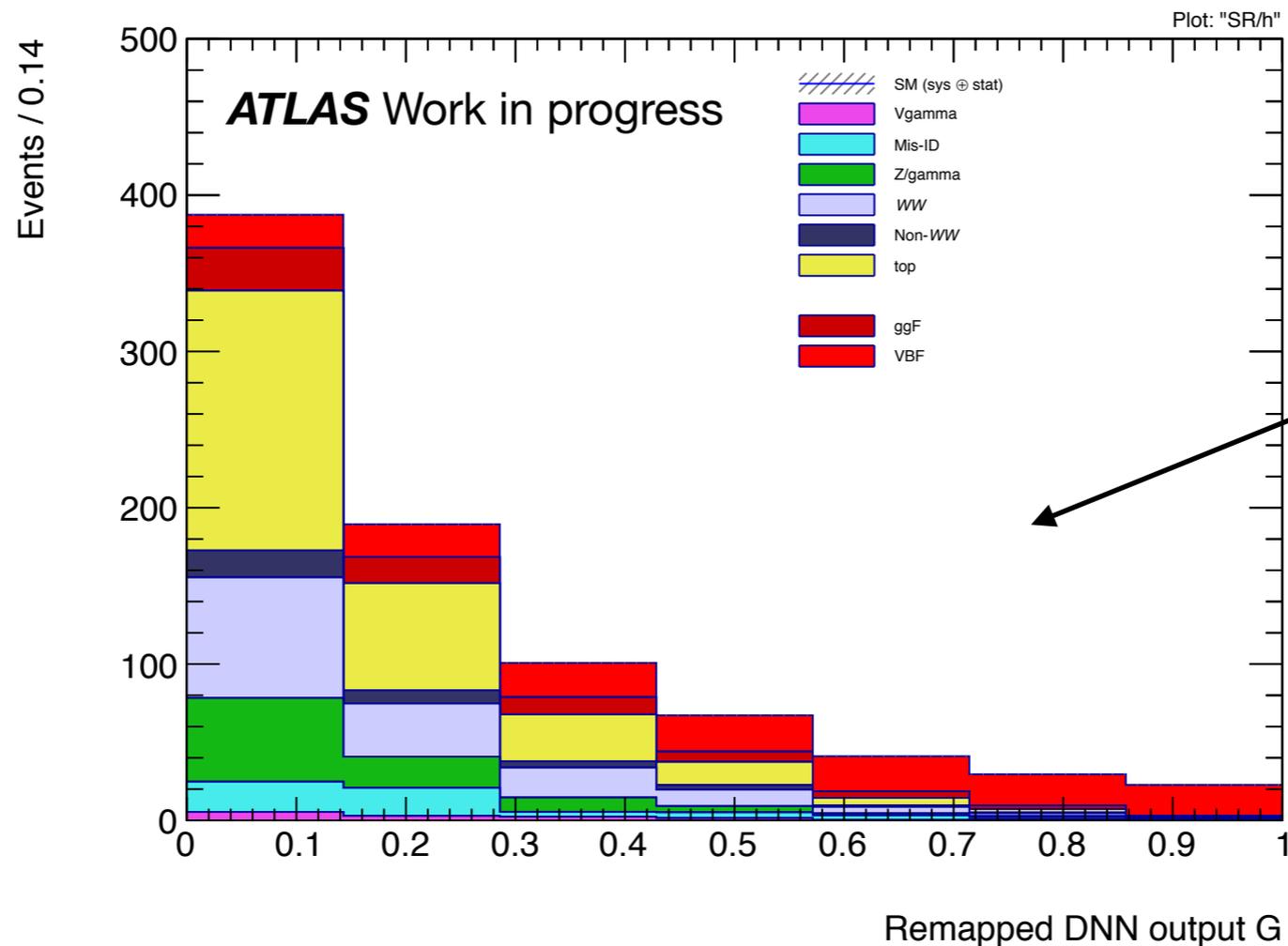


- Methods perform very similarly
- Choose exact binning in final stages of analysis

Input to Final Statistical Analysis

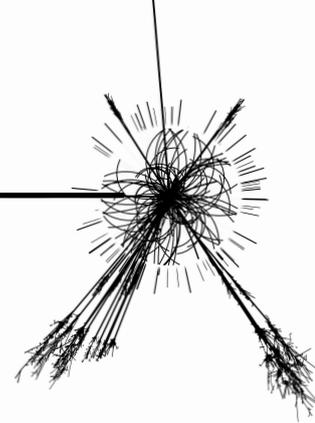


- ▶ After the rebinning the distribution is used in a maximum likelihood fit to and extract parameters of interest (e.g. the cross-section of VBF HWW)



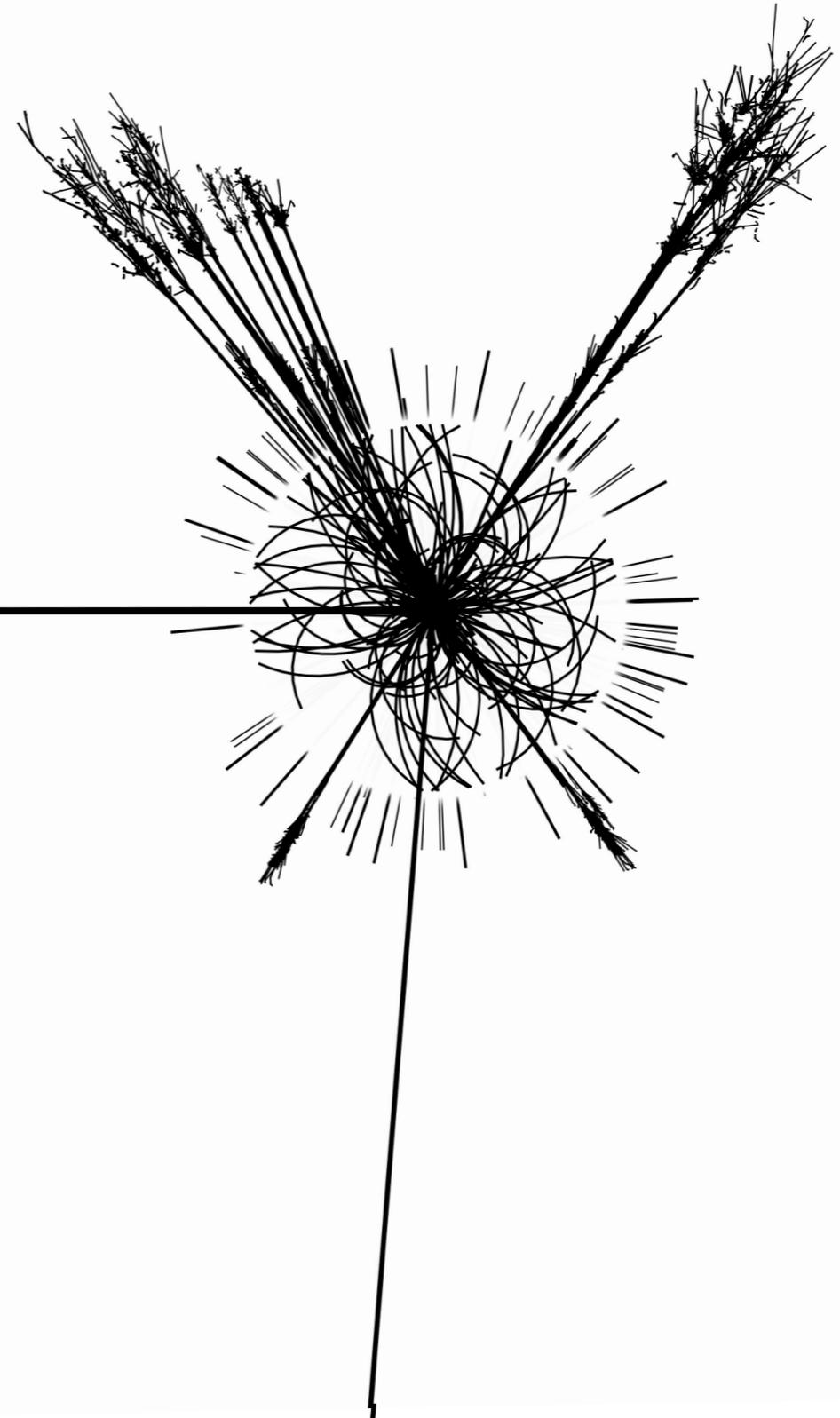
Only “expected” values from simulation so far

- ▶ With the new neural network and rebinning the **expected significance** to observe the VBF HWW process was improved from $\sim 6\sigma$ to $\sim 8\sigma!$ ($p\text{-value} < 10^{-30}$)

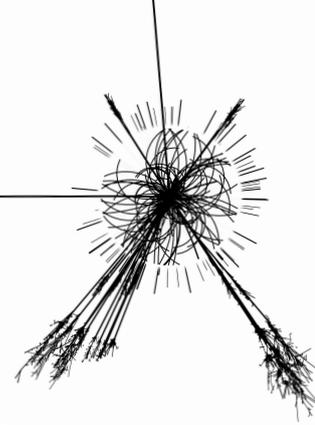


Thanks for the attention!
Questions?

Backup

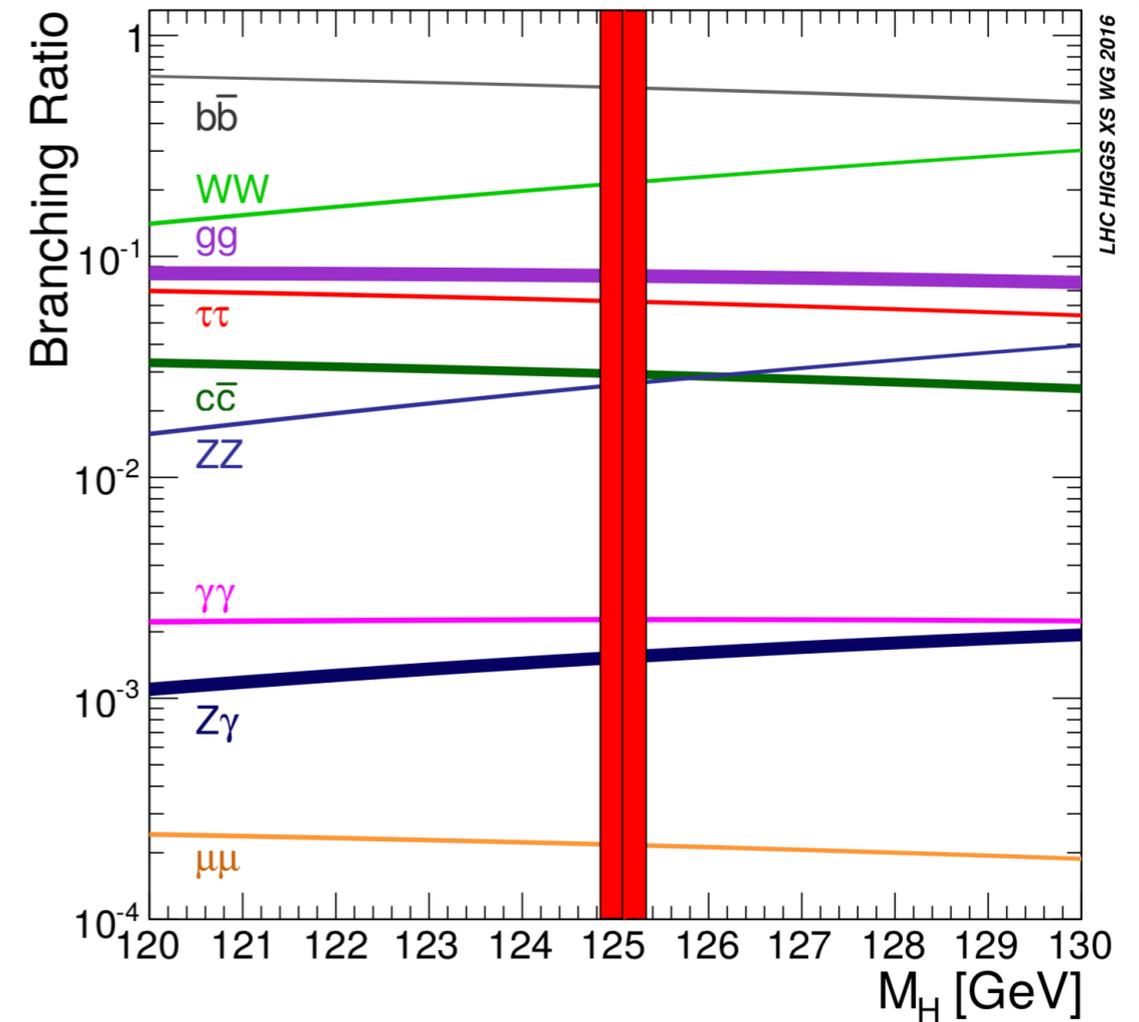
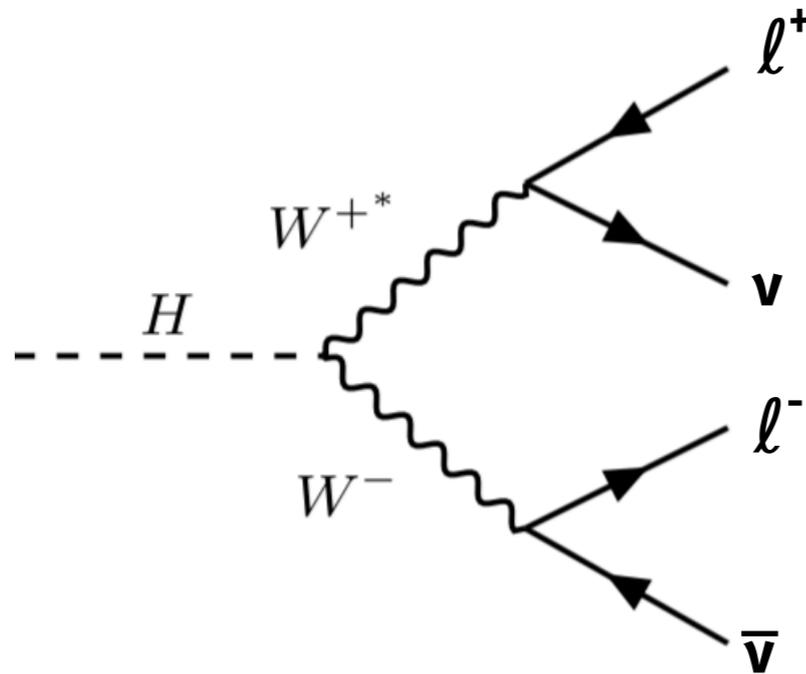


Why Higgs to WW?



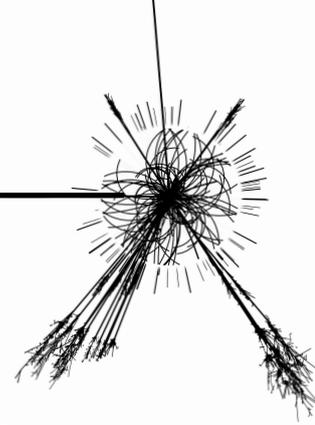
- ▶ Higgs boson has multiple decay channels

- ▶ **Higgs to WW**



- ▶ 2nd largest branching ratio
- ▶ Most sensitive measurement of Higgs to vector-boson coupling
- ▶ Channel needed for conservation of unitarity in WW scattering

Binning Optimization Method 2

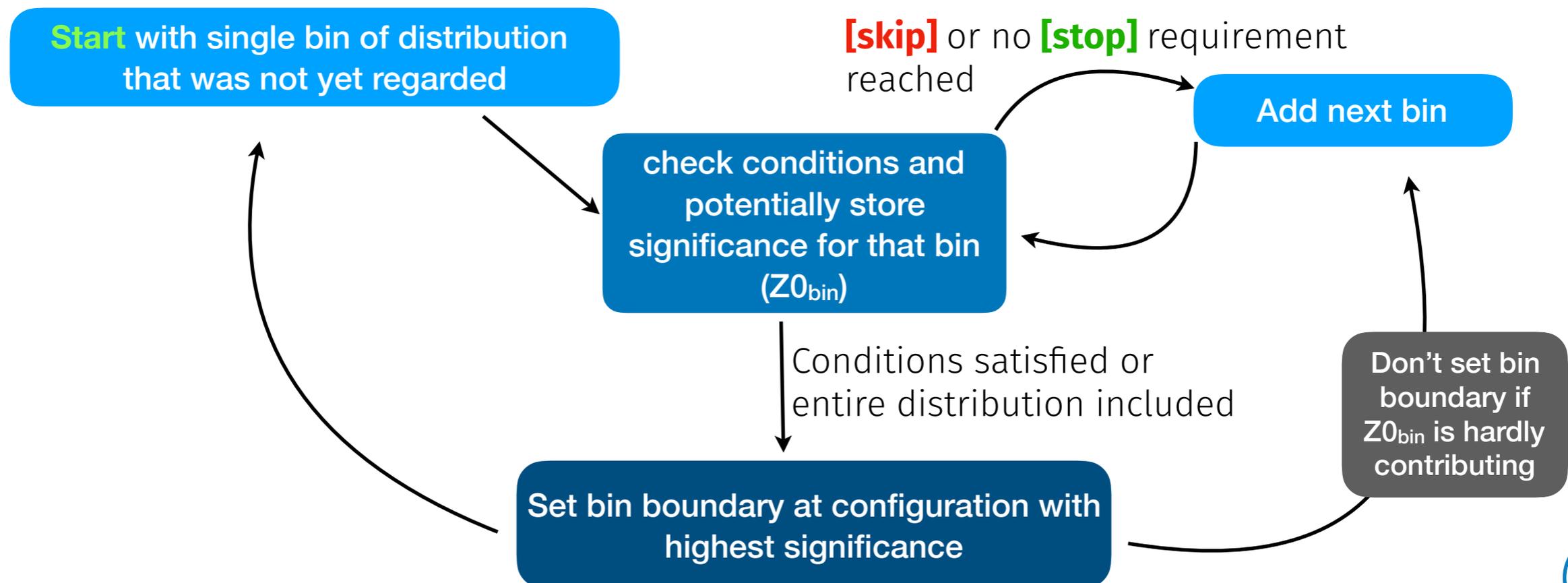


► Method 2: **Significance optimization rebinning**

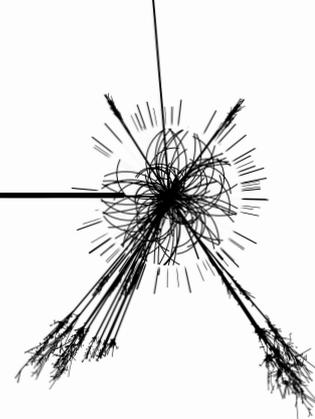
► Inputs (some are optional):

1. minSignal = 10; minimum signal per bin **[skip]**
2. minBkg = 10; minimum bkg per bin **[skip]**
3. maxBkgUnc = 0.2; maximum relative bkg MC uncertainty **[skip]**
4. maxSignal = 30; sufficient signal to set boundary **[stop]**

► Strategy:

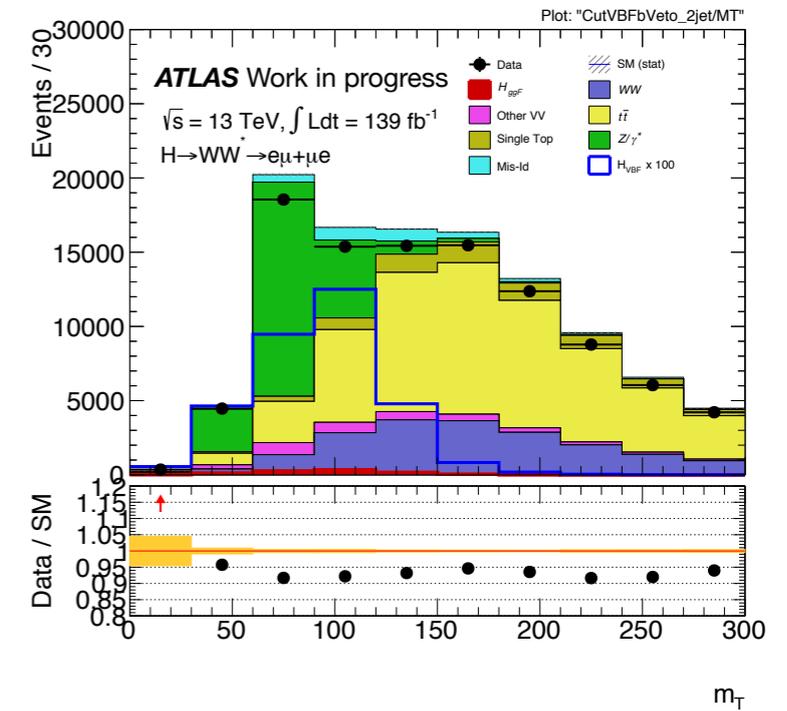
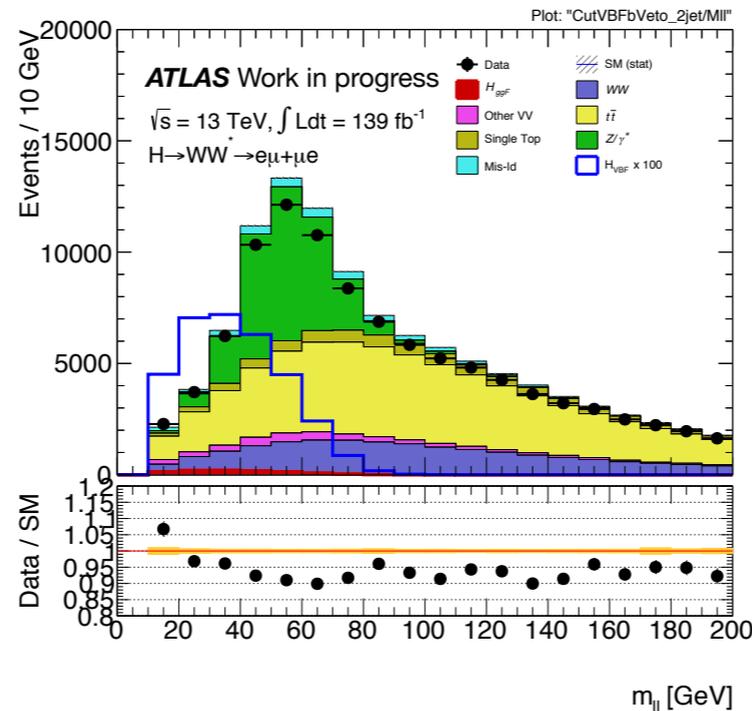
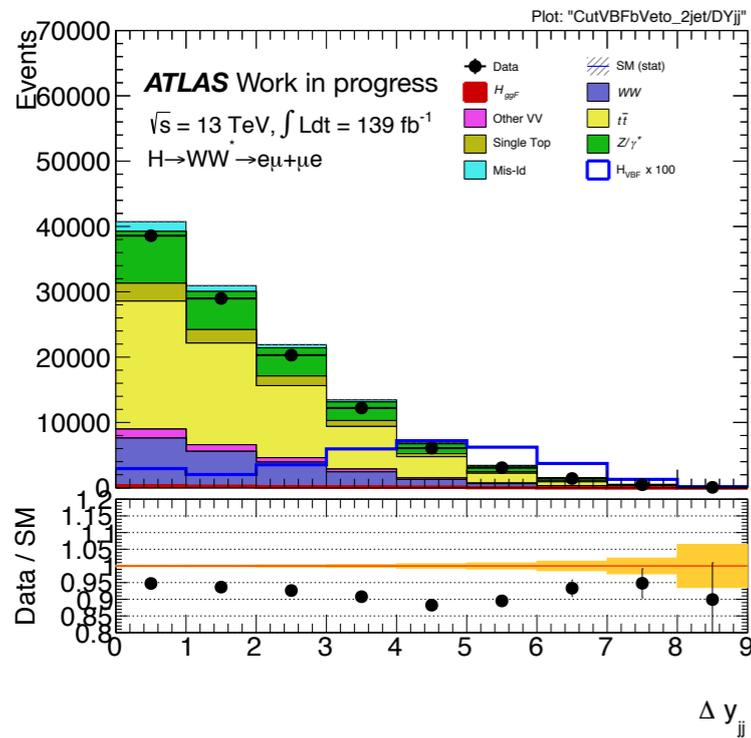


Train a Neural Network for Event Classification

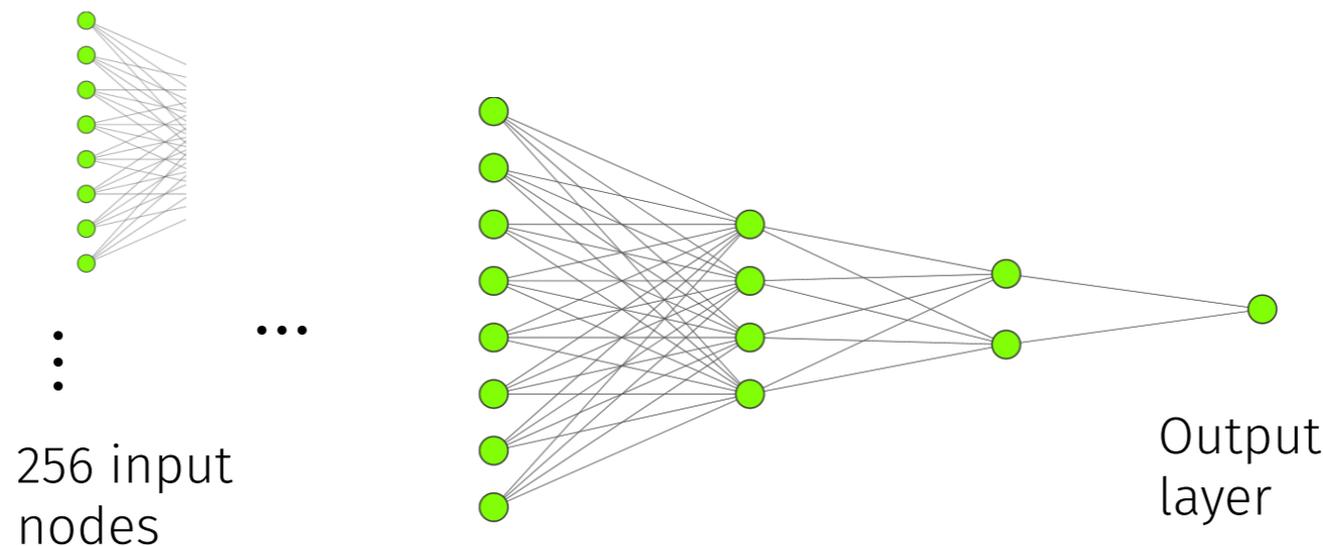


► Variables used in the training (15 in total)

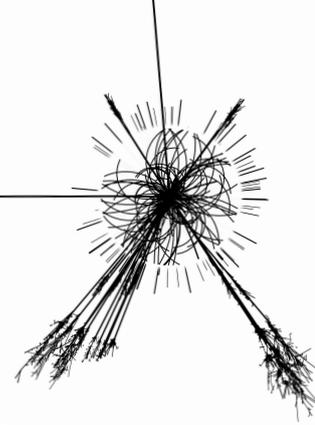
DPHill, DYjj, mjj, mll, mT, ptTot, sumOfCentralitiesL, mL*J* (4), ptJ1/2/3, **METSig** (15)



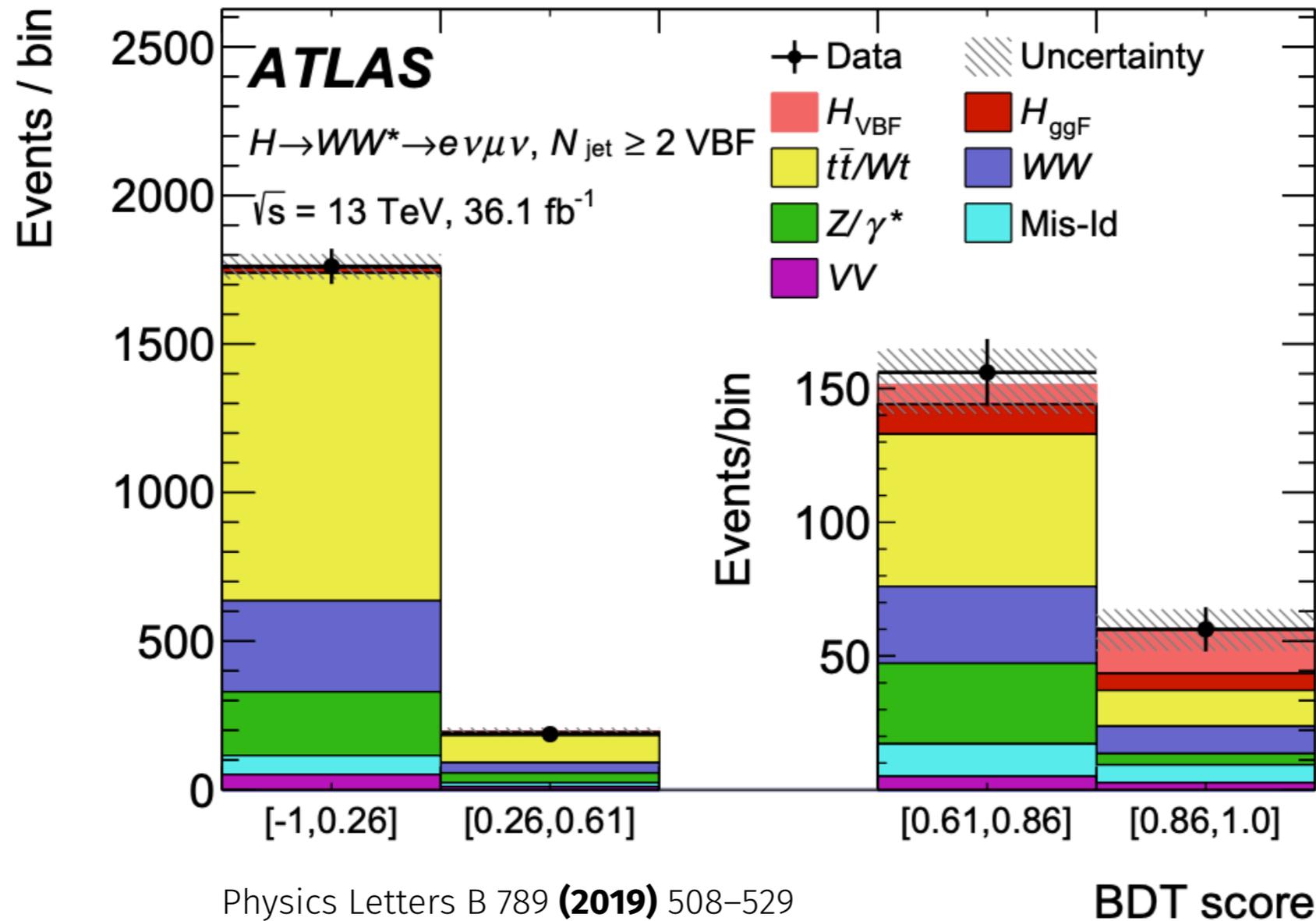
► New "deep" network architecture



“A Counting Experiment”

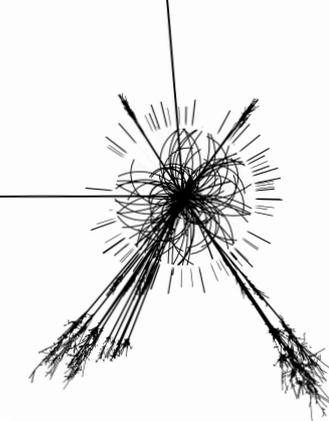


- ▶ VBF HWW results with LHC data from 2015-2016 (Expected significance: 2.6σ)

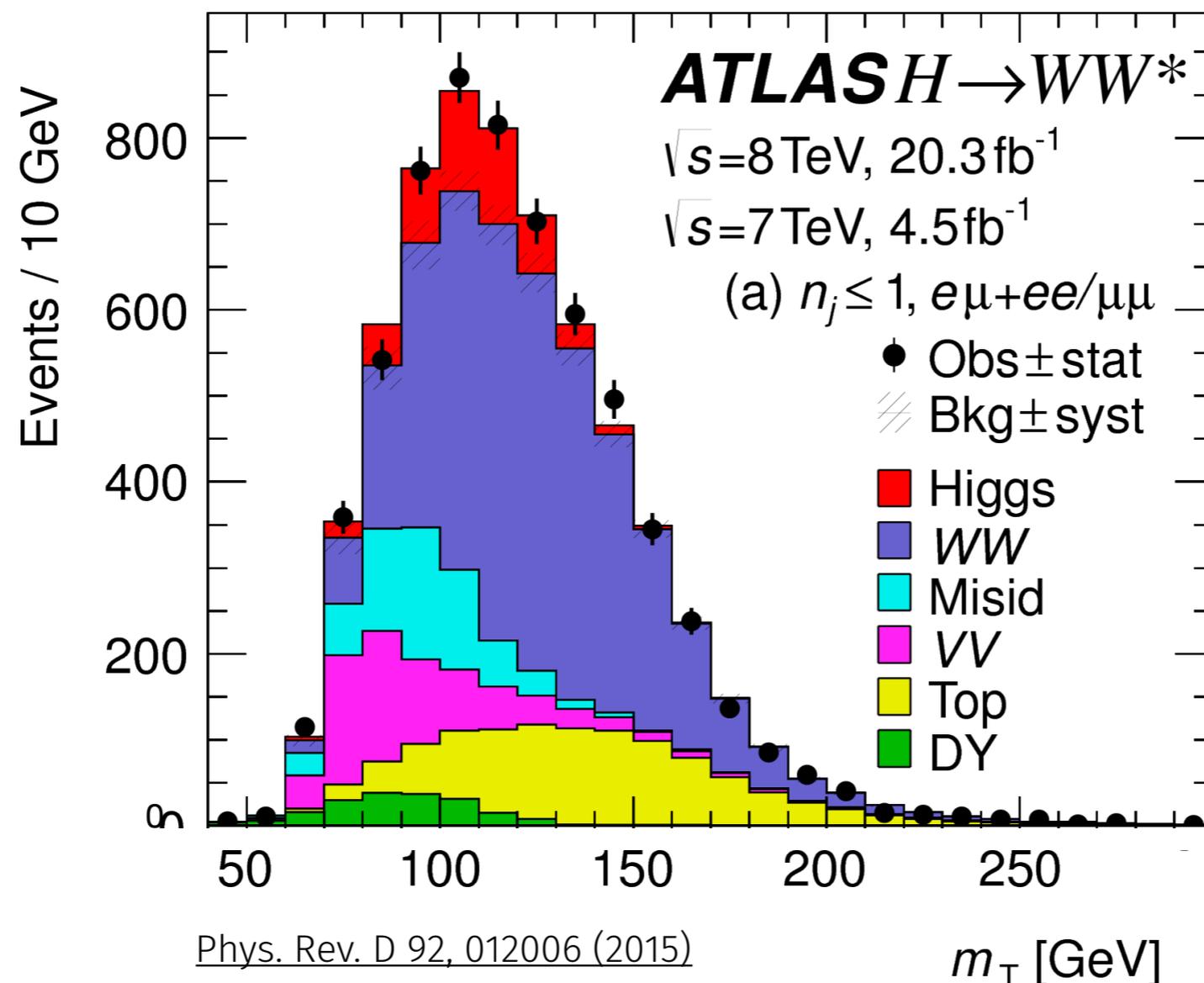


- ▶ Current efforts: precisely measure **properties** of Higgs boson with data recorded in 2015-2018. (side-goal: measure exclusive VBF HWW with significance $> 5\sigma$)

“A Counting Experiment”



- ▶ Results contributing to the Higgs boson discovery



- ▶ Current efforts: precisely measure **properties** of Higgs boson (with data recorded in 2015-2018)