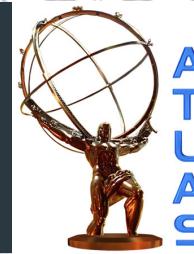


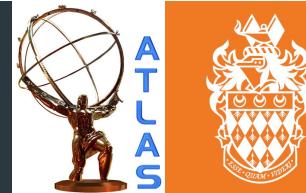
# Application of the matrix element method in the search for ttH, $H \rightarrow bb$ in the lepton+jets channel, using the ATLAS detector

Ian Connelly, on behalf of the ATLAS Collaboration  
Higgs Couplings 2015, Durham  
12 - 15 October 2015



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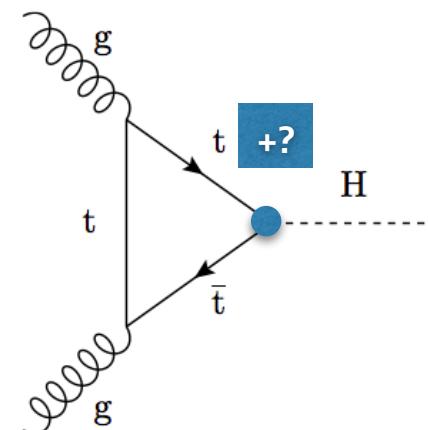
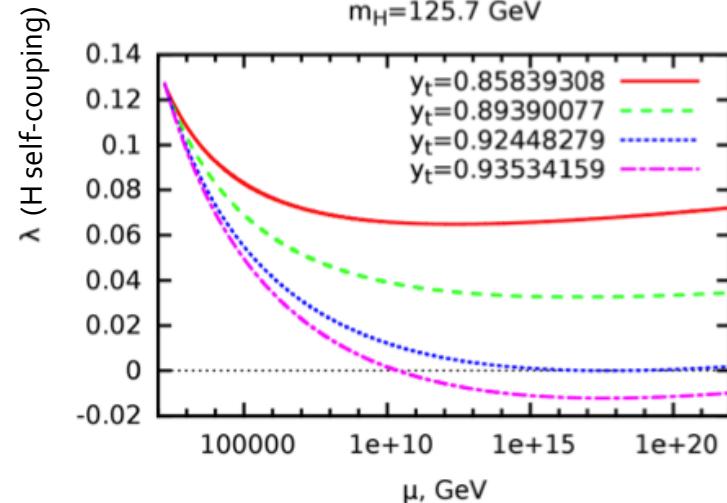
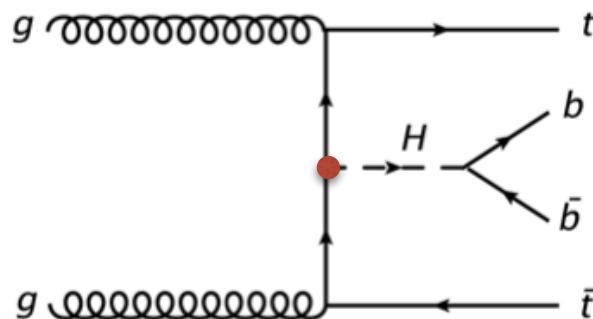
# Theoretical Motivation



[J.Exp.Theor.Phys. 120 (2015) 3, 335-343; ZhETF 147 (2015) 3, 389]

## Why ttH?

- Measurement of Higgs properties.
- The top Yukawa coupling ( $Y_t$ ) is approximately unity.
  - EW vacuum stability in SM.
  - Probing this value can provide information of the scale of New Physics.
- This search provides a direct measurement of  $Y_t$ .
  - Leading order tree diagram.
  - $\sigma_{ttH} \propto |Y_t|^2$
  - Required to constrain NP contributions in indirect searches ( $ggF, H \rightarrow \gamma\gamma$ ) in a model-independent way.



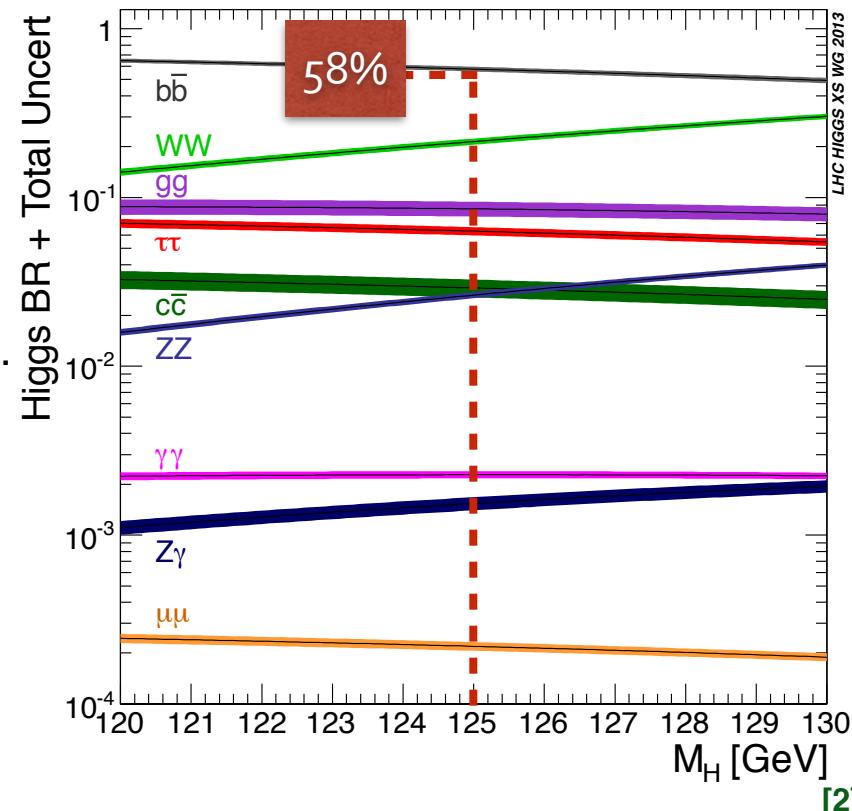
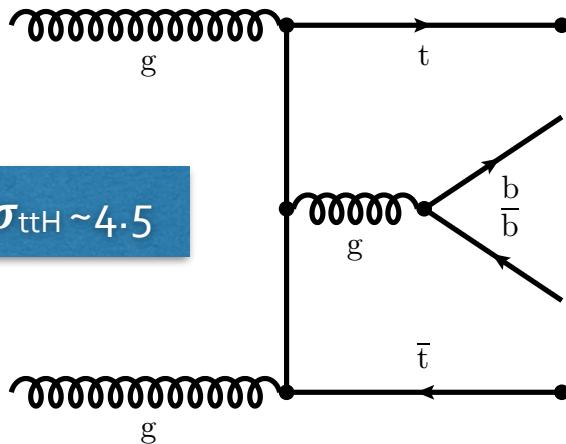
# Theoretical Motivation



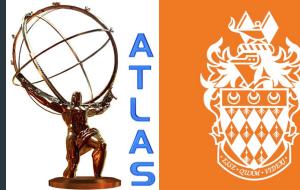
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## Why $H \rightarrow bb$ ?

- $t\bar{t}H$  is largely suppressed at the LHC compared to alternative production modes
  - $129.3 \text{ fb}$  @ 8 TeV (5.8% of total Higgs cross-section).
  - Expected  $\sim 2600$  events in 8 TeV dataset.
- Largest BR of  $m_H = 125 \text{ GeV}$  Higgs is  $H \rightarrow bb$ 
  - Provides the most events for single decay signature.
- $t\bar{t}+X$  provides dominant background processes.
  - $\sim 2000$  times larger cross-section.

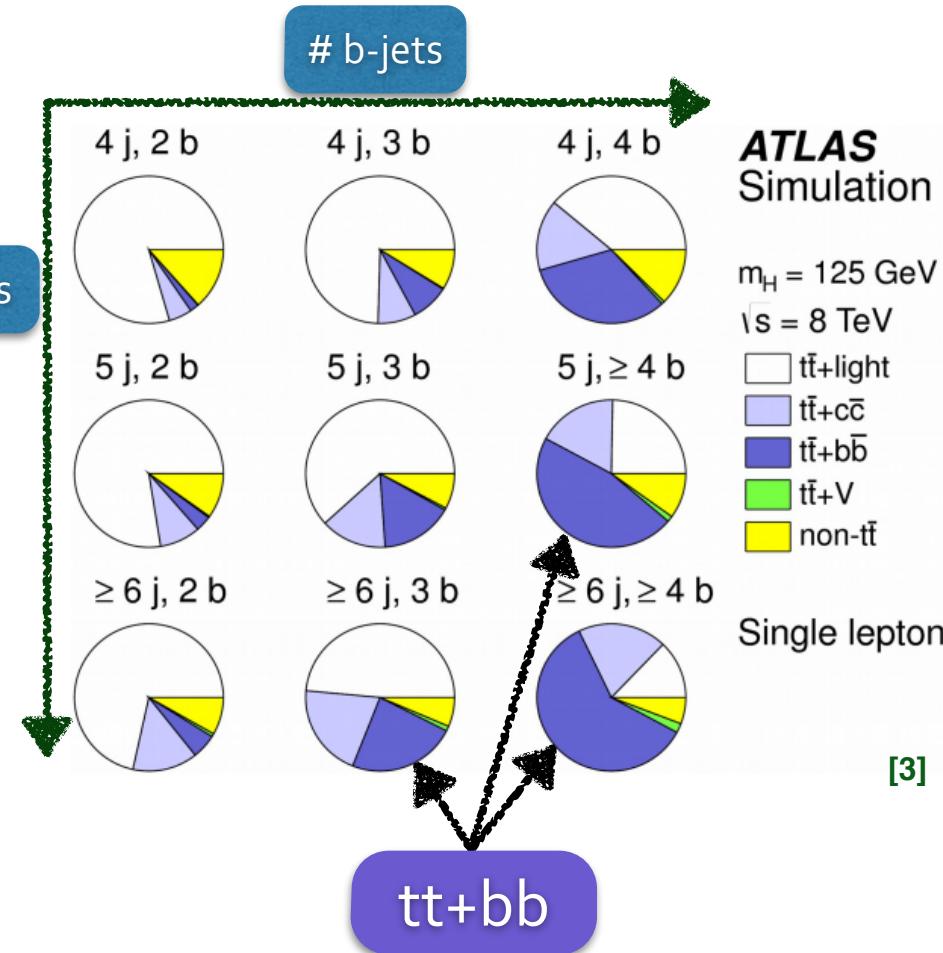
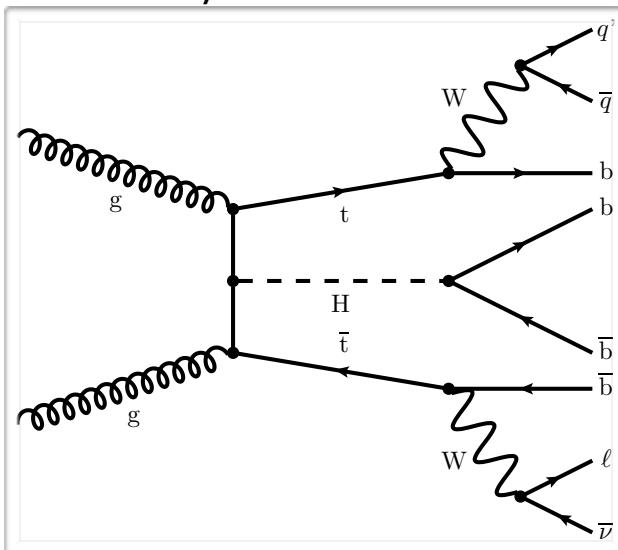


# Analysis Overview



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- Event selection
  - 1 isolated lepton ( $pT > 25 \text{ GeV}$ )
  - at least 4 jets ( $pT > 25 \text{ GeV}$ )
  - at least 2 b-tagged jets
- Events are categorised by the number of jets and the number of b-jets.
  - Different background compositions.
  - Profiled likelihood fit to data used to contain systematic uncertainties.

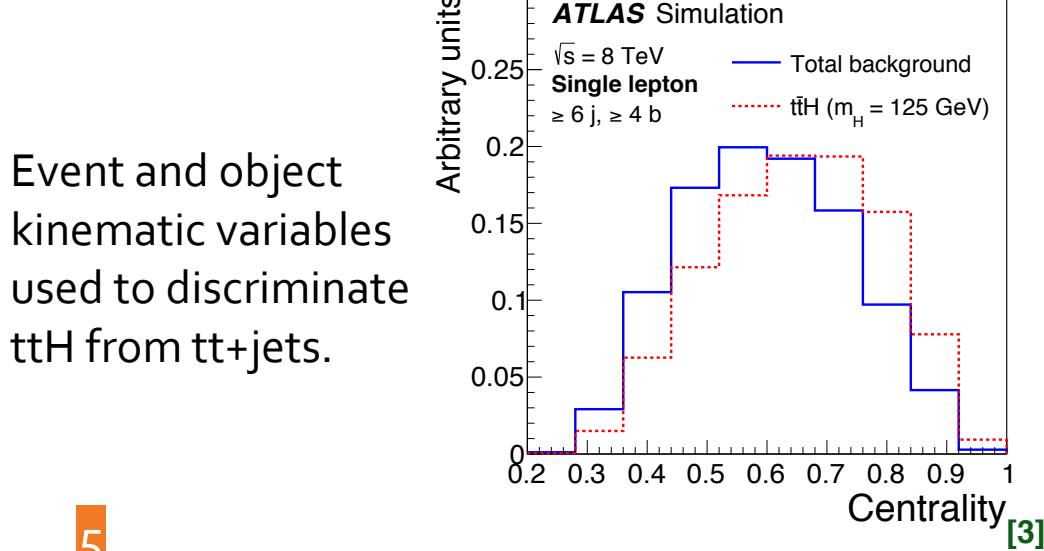


# Analysis Overview

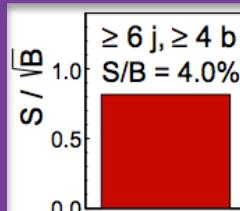
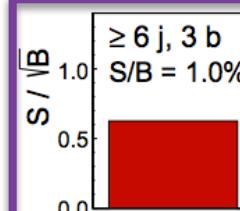
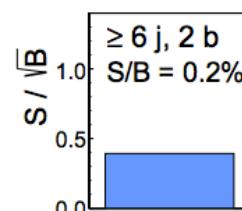
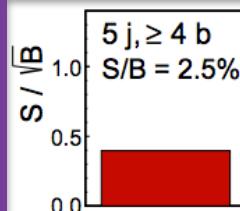
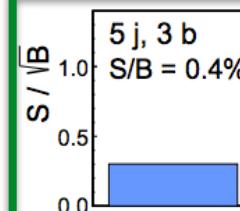
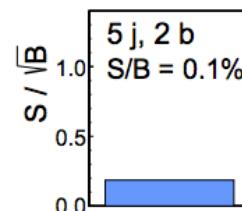
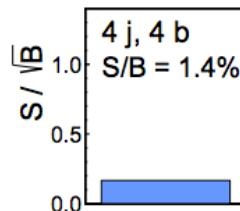
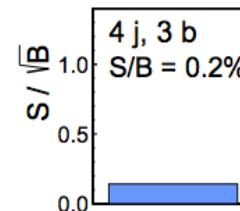
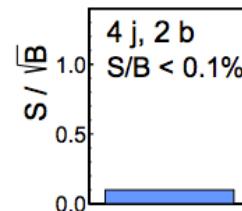


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- **Signal-rich** regions
  - $S/B > 1\%$  and  $S/\sqrt{B} > 0.3$
  - Large  $t\bar{t}+bb$  background in these regions
- **Neural networks** trained to separate signal and background processes.
  - **Additional network** used to separate  $t\bar{t}$ +heavy flavour from  $t\bar{t}$ +light flavour.

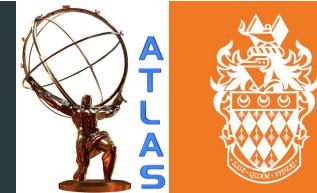


**ATLAS Simulation**  
 $\sqrt{s} = 8 \text{ TeV}, 20.3 \text{ fb}^{-1}$

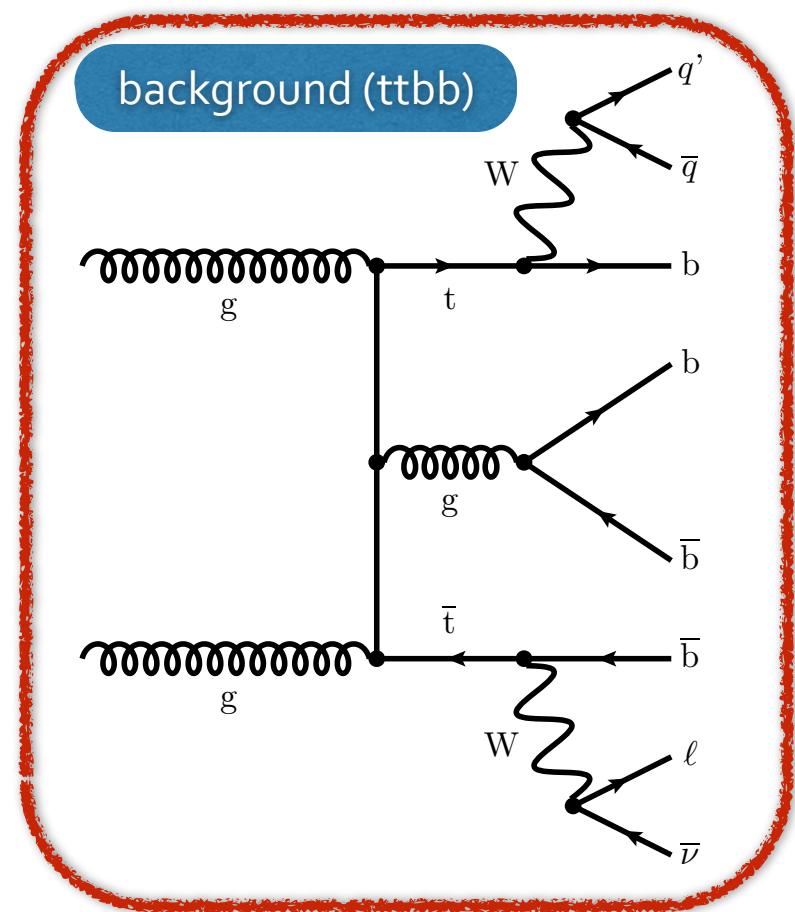
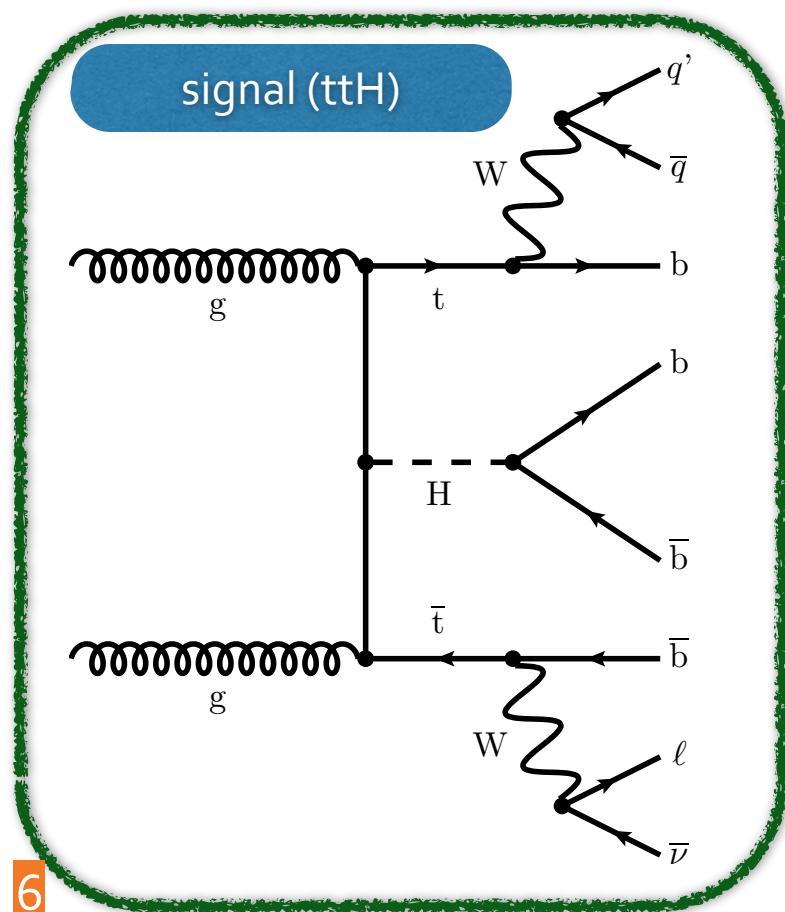


[3]

# The Matrix Element Method



- Determines the probability of the data under a hypothesis.
  - Test two hypotheses ( $H$ ) : ttH (signal) and ttbb (background).



# The Matrix Element Method



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- Determines the probability of the data under a hypothesis.
  - Test two hypotheses ( $H$ ) : ttH (signal) and ttbb (background).
- The matrix element method links a set of measured quantities ( $\mathbf{x}$ , e.g. b-jet energy) with a set of unobservable partonic objects ( $\mathbf{y}$ , e.g. b-quark energy) associated to a hypothesised process.
  - $\boldsymbol{\alpha}$  parameterises the theoretical model.

$$P(\mathbf{x}|H, \boldsymbol{\alpha}) = \frac{(2\pi)^4}{\sigma_i^{\exp(\boldsymbol{\alpha})}} \int d\mathbf{p}_A(\mathbf{y})d\mathbf{p}_B(\mathbf{y})f(p_A)f(p_B) \frac{|\mathcal{M}(\mathbf{y}|H, \boldsymbol{\alpha})|^2}{\mathcal{F}} W(\mathbf{x}|\mathbf{y})d\Phi_N(\mathbf{y})$$

Normalisation

PDFs

Transition matrix element

Transfer functions (TF)

Phase space volume element

# Integration Strategy



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$$P(\mathbf{x}|H, \boldsymbol{\alpha}) = \frac{(2\pi)^4}{\sigma_i^{\exp(\boldsymbol{\alpha})}} \int dp_A(\mathbf{y}) dp_B(\mathbf{y}) f(p_A) f(p_B) \frac{|\mathcal{M}(\mathbf{y}|H, \boldsymbol{\alpha})|^2}{\mathcal{F}} \mathcal{W}(\mathbf{x}|\mathbf{y}) d\Phi_N(\mathbf{y})$$

2 partons

4 b-quarks  
2 light quarks  
1 lepton  
1 neutrino

- Each object contributes three degrees of freedom.
  - Nominally 30-dimensional integration
- Approximations used to reduce this to an 6-dimensional integration.
  - Analytic initial state kinematics.
  - Jet directions are well-measured ( $\text{TF} = \delta$  function).
  - Lepton four-vectors are well-measured ( $\text{TF} = \delta$  function).
  - MET interpreted as neutrino  $p_T$  measurement.

# Integration Strategy



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$$P(\mathbf{x}|H, \boldsymbol{\alpha}) = \frac{(2\pi)^4}{\sigma_i^{\exp(\boldsymbol{\alpha})}} \int dp_A(\mathbf{y})dp_B(\mathbf{y})f(p_A)f(p_B) \frac{|\mathcal{M}(\mathbf{y}|H, \boldsymbol{\alpha})|^2}{\mathcal{F}} \mathcal{W}(\mathbf{x}|\mathbf{y})d\Phi_N(\mathbf{y})$$

- Integration strategy :

Dijet pair	bb mass	leading b-quark energy
Top decays	b-quark energy	b-quark energy
Hadronic W	W mass	leading light-quark
Leptonic W	Neutrino p	

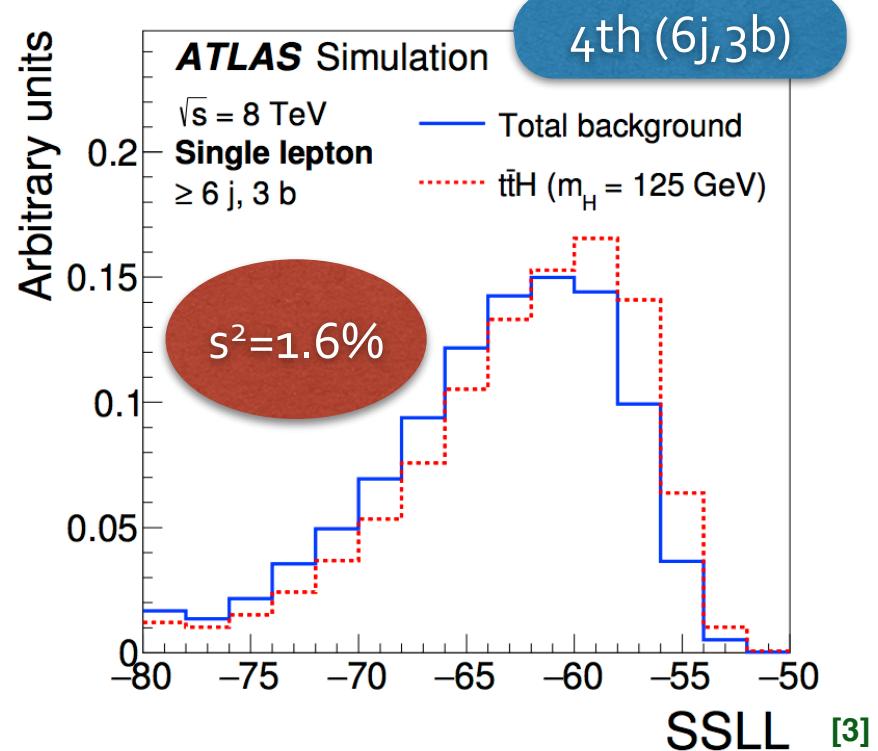
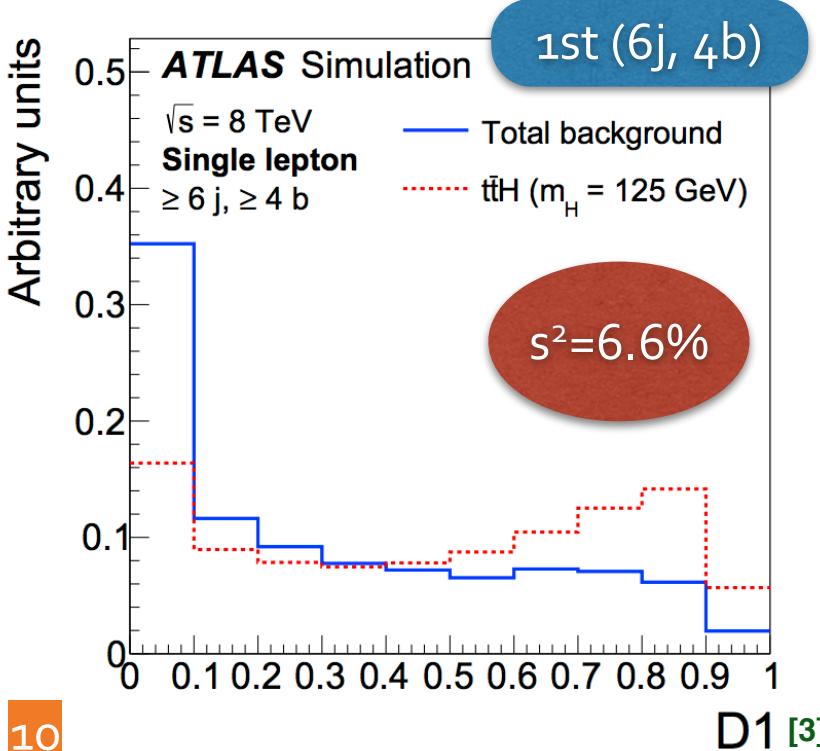
- Indistinguishable objects in  $\mathbf{x}$  are permuted with their counterparts in  $\mathbf{y}$ .
  - $\geq 6j, \geq 4b - 12$  permutations.
  - $\geq 6j, 3b - 36$  permutations.
- Per-event integration time :  $O(1-2$  minutes).
  - Feasible for large, complex analysis such as ttH.

# The Matrix Element Method



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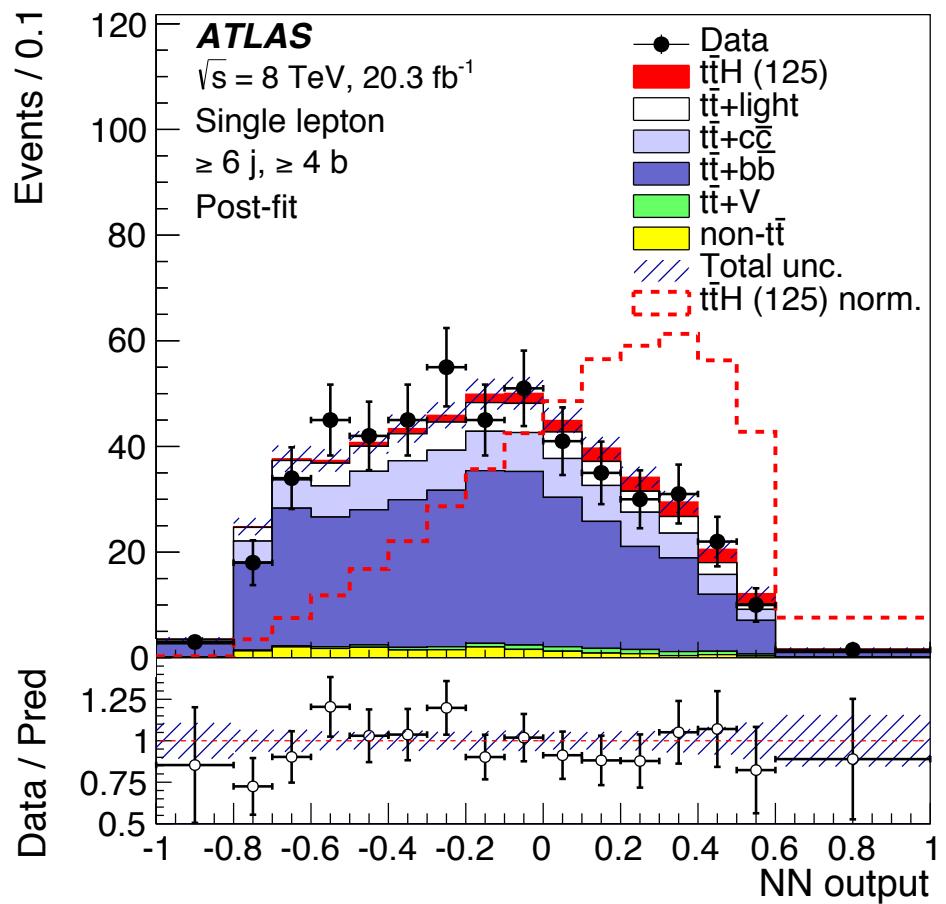
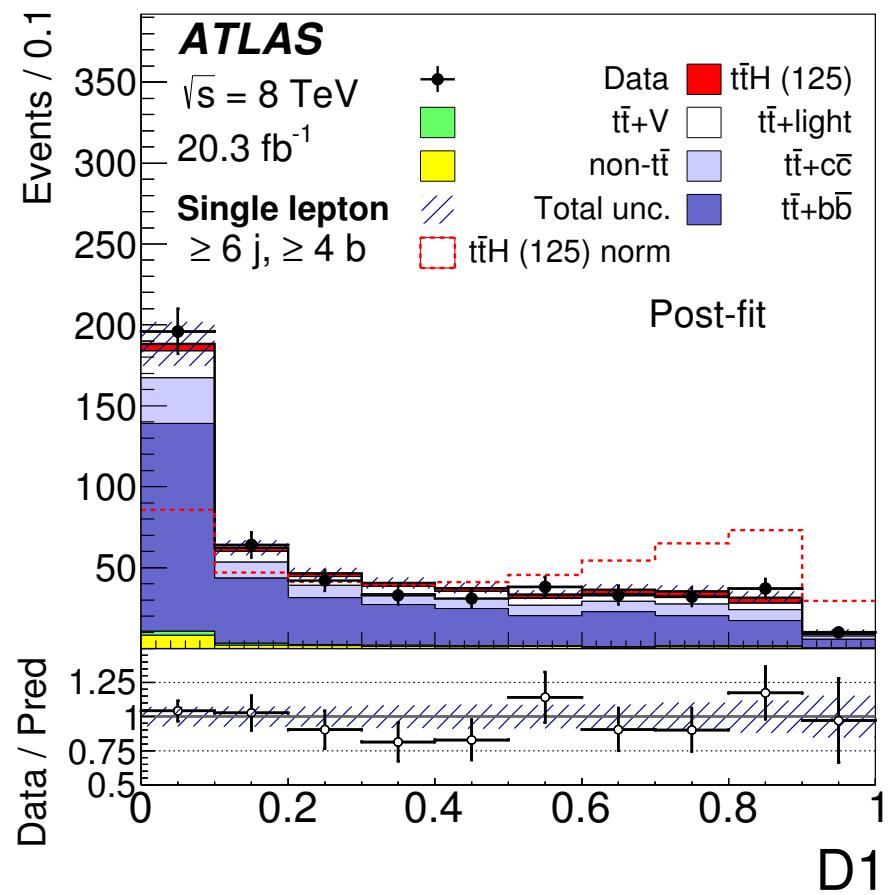
- Two variables constructed using matrix element method.
  - **D<sub>1</sub>** : Monotonic Neyman-Pearson likelihood ratio.
  - **SSLL** : Logarithm of  $\Sigma P(x|S)$ .
- Included in the set of neural network input variables.
  - Very discriminating and ranked highly by NN training.



# Post-Fit Neural Network



Well-modelled MEM variables and final NN output after fitting to data.



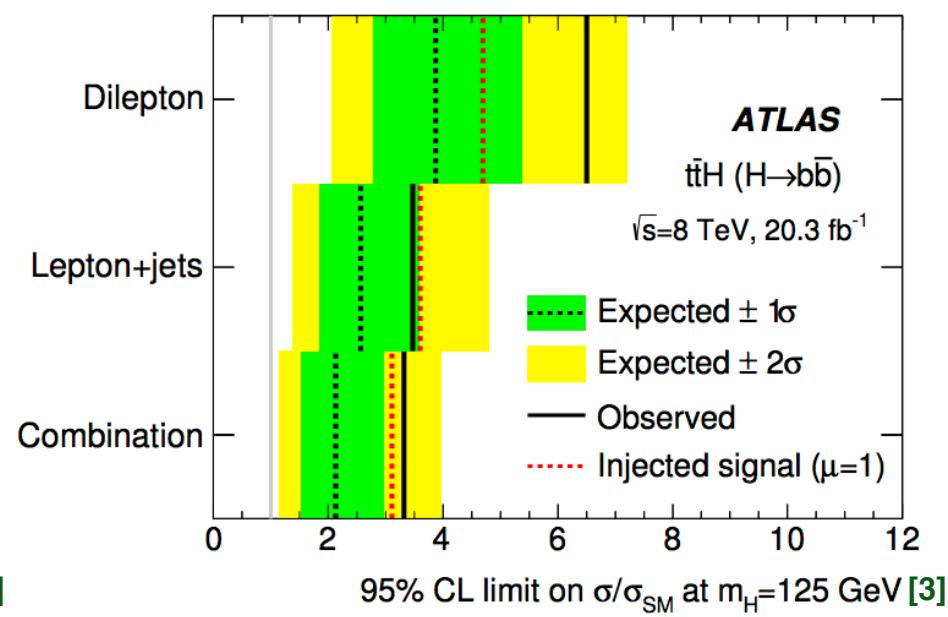
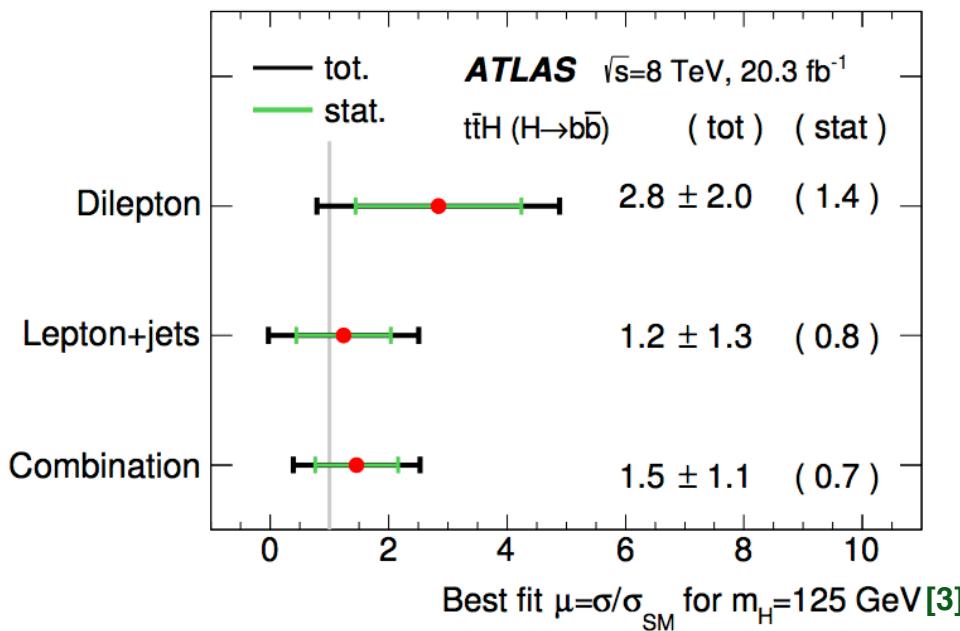
# Analysis Results



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- Statistical combination of the lepton+jets and dilepton analyses.
  - Approaching SM sensitivity.

$$\mu = \sigma/\sigma_{SM}$$



# Summary



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- Successful Run 1 results
  - $\mu_{\text{obs}} = 1.5 \pm 1.1$ .
  - Exclude  $\mu > 3.4$  (2.2 expected) at 95% CL<sub>s</sub>.
  - Included in ATLAS Higgs coupling combination.
- A highly complex analysis has been carried out in the search for ttH(H → bb).
  - Many advanced data-analysis techniques used - focussed here on MEM.
    - Success of the analysis highly dependent on modelling ttbar processes.
  - Statistical combination of single lepton and dilepton channels.
  - Inclusion of MEM improved expected limit by **16%**.
    - Studies indicate similar performance gain possible in dilepton analysis.
- Looking at Run 2
  - Enough data to make an observation?
    - Sophisticated techniques like MEM required to maximise performance.
      - Work is on-going for MEM in Run 2.
  - Hopefully reach SM sensitivity!

## References



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- (1) [J.Exp.Theor.Phys. **120** (2015) 3, 335-343; ZhETF **147** (2015) 3, 389]
- (2) [LHCHXSWG]
- (3) [Eur. Phys. J. C (2015) **75**:349]
- (4) [ATLAS-CONF-2014-011]

# Backup



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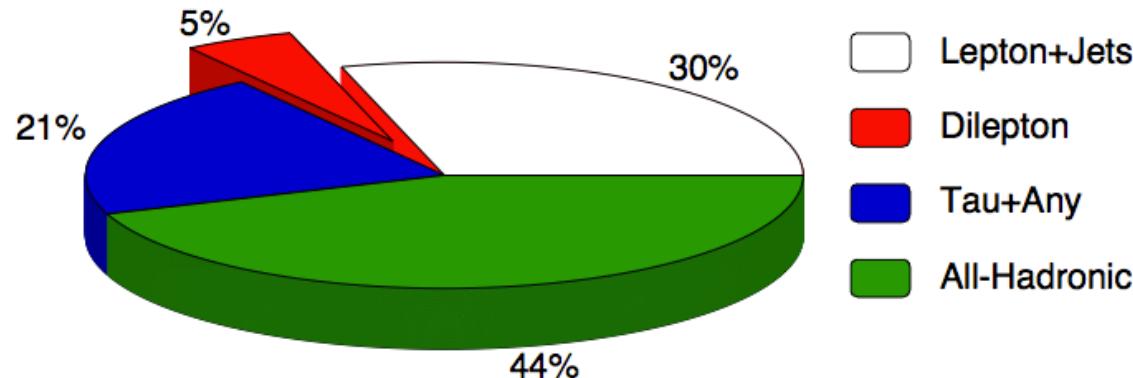
# Theoretical Motivation



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$t\bar{t}H$  events can also be categorised by the top-quark pair decay.

- Leptonic ( $e/\mu$ ) decays provide clean/ well-understood trigger objects.
  - $\sim 450$  expected signal events in 8 TeV dataset.





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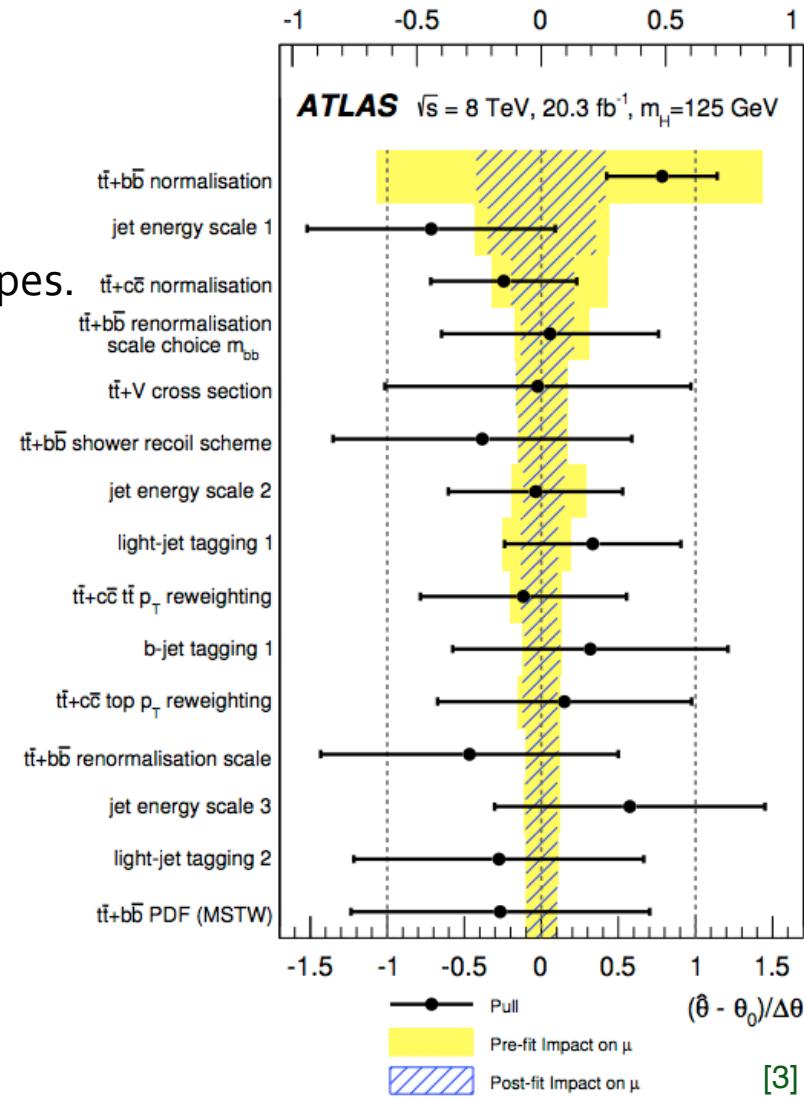
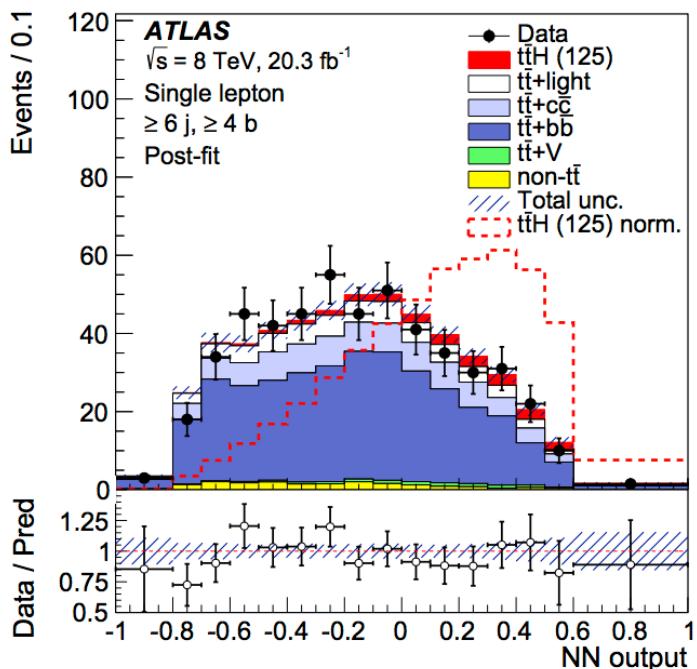
- MEMTool
  - Modular design to allow user flexibility
  - ME generated by MadGraph 5 (standalone, C++)
  - Numerical integration performed by VEGAS (GSL 1.16)
  - PDFs from LHAPDF 5.9.1 (CTEQ 6 parameterisation)

# Simultaneous Fit



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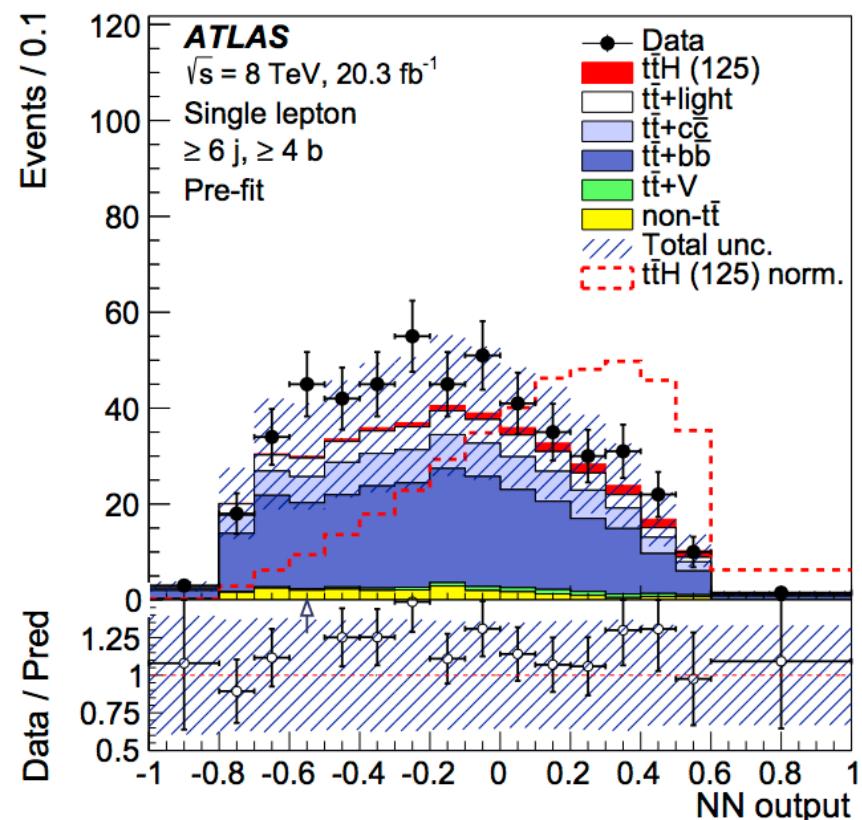
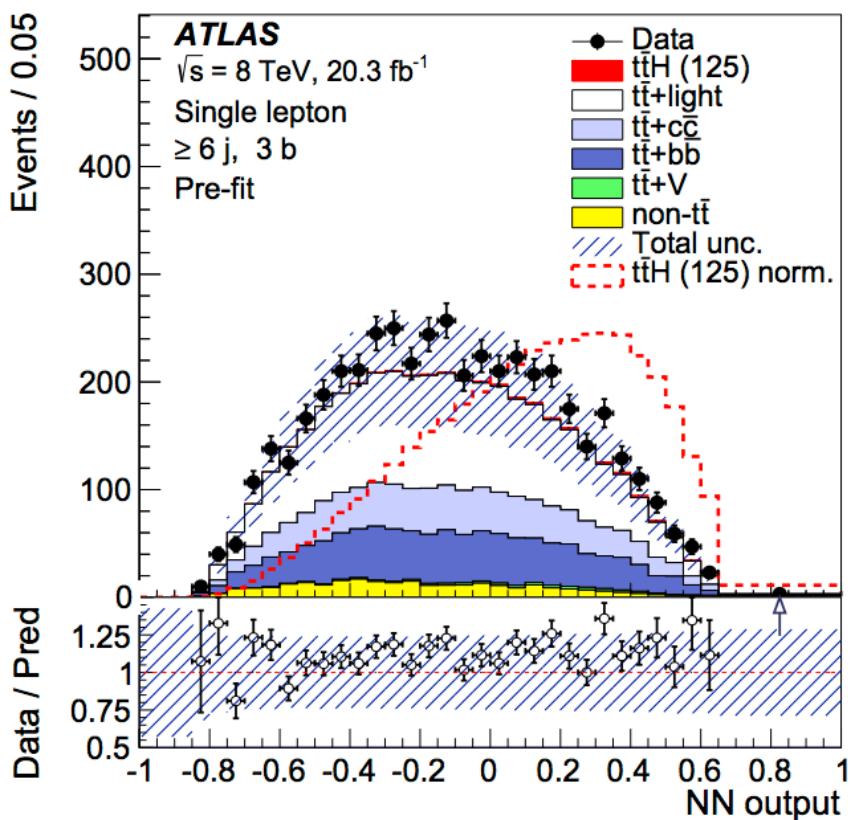
- Binned profiled likelihood fit used to constrain systematic uncertainties ( $\theta$ ).
- Parameter of interest :  $\mu = \sigma_{\text{obs}} / \sigma_{\text{ttH}}$
- Simultaneous fit over all analysis categories.
  - Requires good modelling of background shapes.
  - Requires background-only regions to fit dominant backgrounds to data.



# Neural Network Output (Pre-fit)



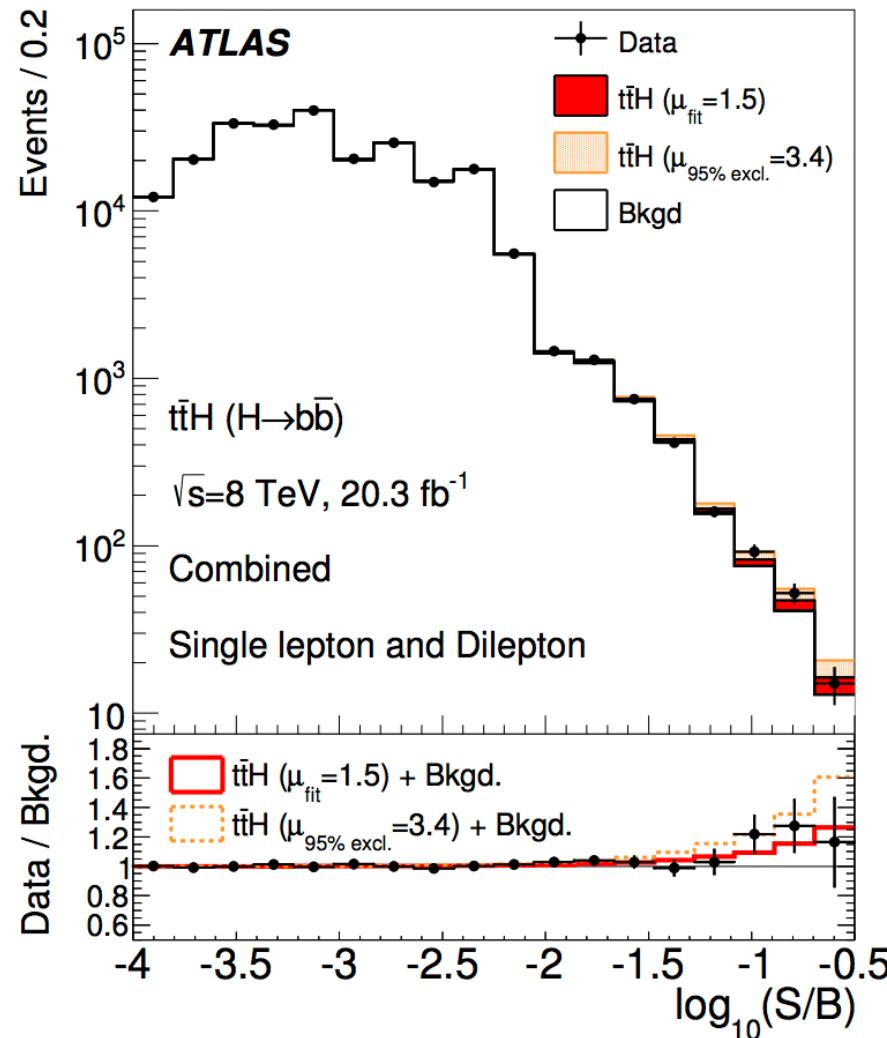
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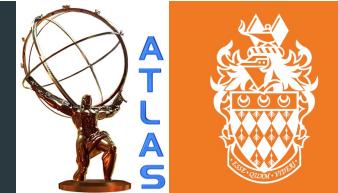
# Analysis Results



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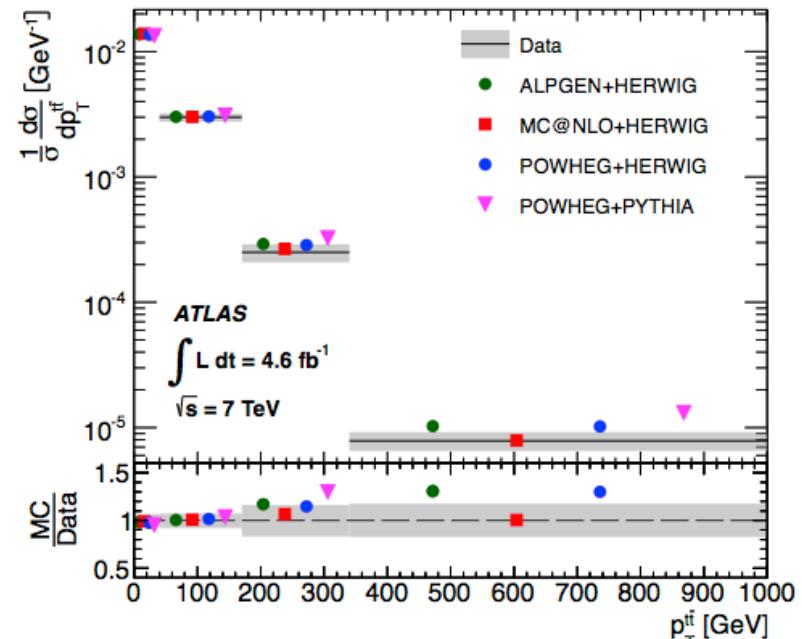
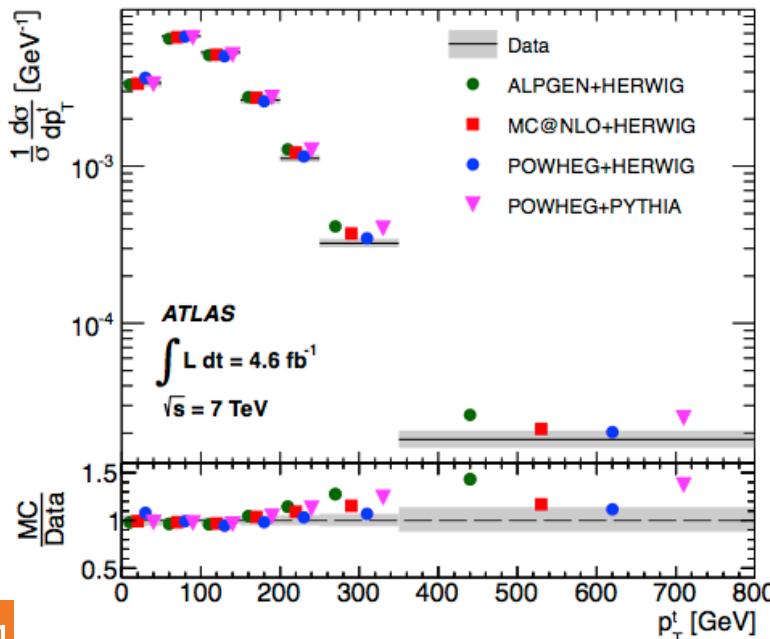


# Signal and Background Modelling



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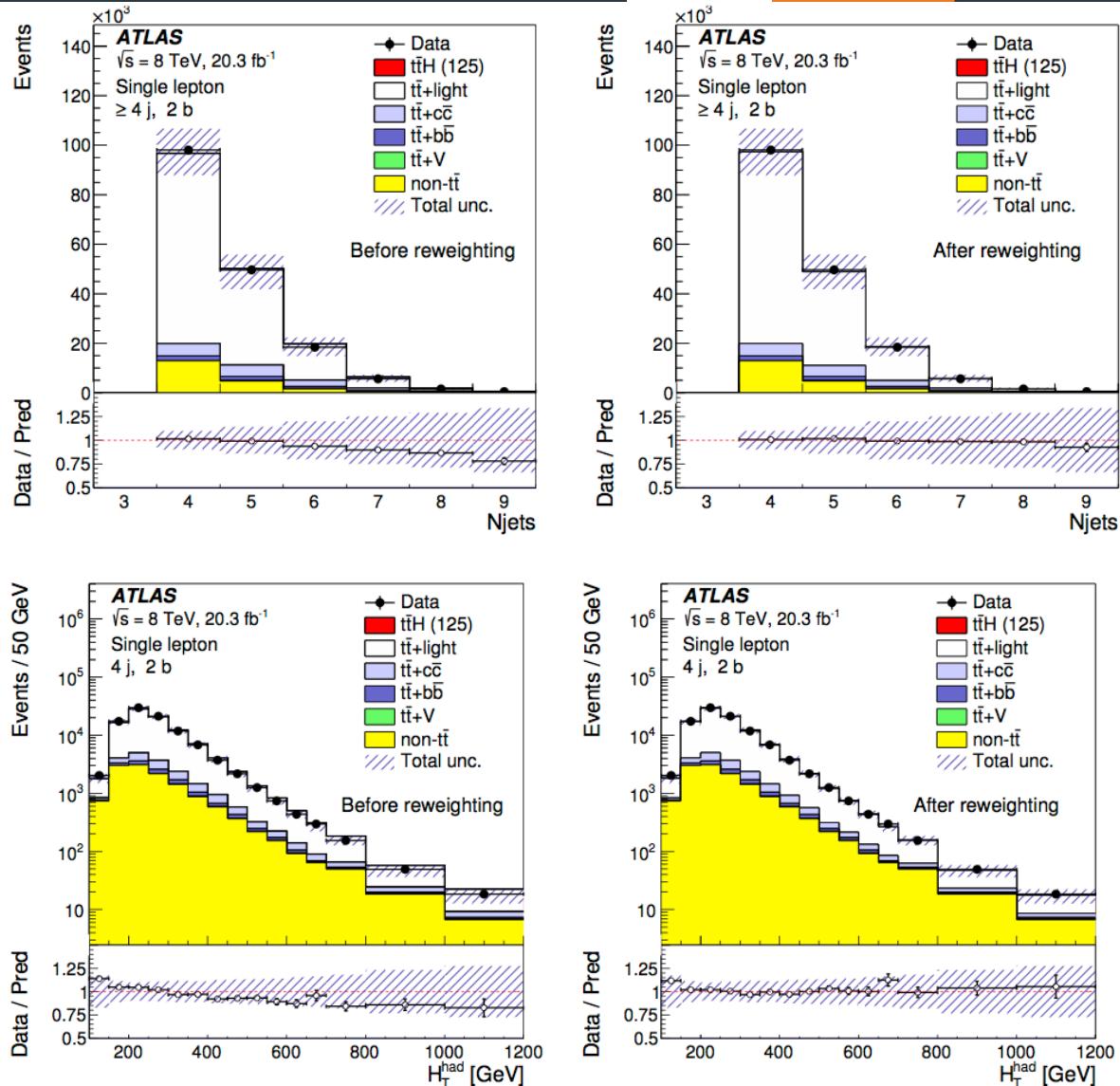
- Important to have the best possible modelling for background processes to prevent any mismodelling being interpreted as signal.
- NLO samples used for signal and dominant backgrounds.
  - tt (inclusive) normalised to NNLO+NNLL (QCD).
  - tt+bb modelling benefitted from recent NLO simulation
- tt sample reweighted to 7 TeV unfolded XS analysis to correct top-quark kinematic modelling.



# Top System Modelling



- Reweighting  $t\bar{t}$   $p_T$  improved number of jets beyond LO diagrams.
- Reweighting top  $p_T$  improved the hadronic summed  $p_T$  distributions.



# Heavy Flavour Modelling ( $t\bar{t}$ +HF)



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- Large difference in  $t\bar{t}$  + unresolved  $g \rightarrow b\bar{b}$  events comparing parton-shower modelling with NLO event generation.

