

# Measurement of the production cross section for the single *top* quark in association with a $Z$ boson at the ATLAS detector

Tesi di Laurea Magistrale di:  
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Relatrice: **Professoressa Lidia Dell'Asta**

Correlatore: **Professore Attilio Andreazza**



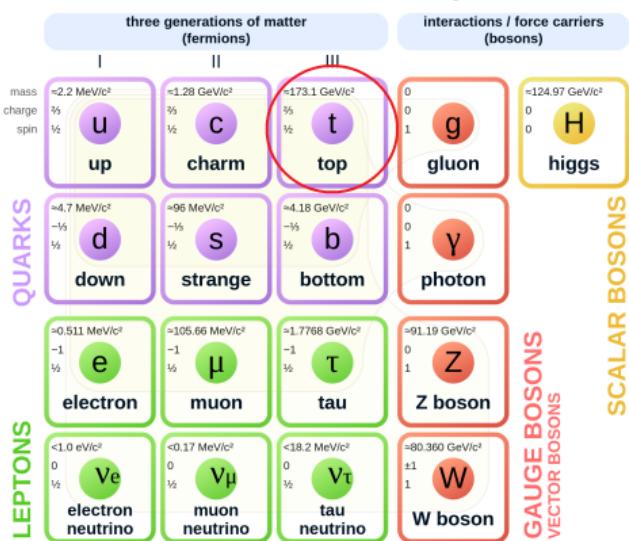
- 1 Theory
- 2  $tZq$  associated production
- 3 Event selection
- 4 MVA training
- 5 Fit
- 6 Ratio of *top* and *antitop* cross sections in  $tZq$  production
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# The Standard Model and the top quark

## Standard Model of Elementary Particles



## Standard Model (SM)

- Gauge theory describing 12 elementary particles, their interactions mediated by 4 vector bosons, and a scalar (Higgs boson)

## Top quark

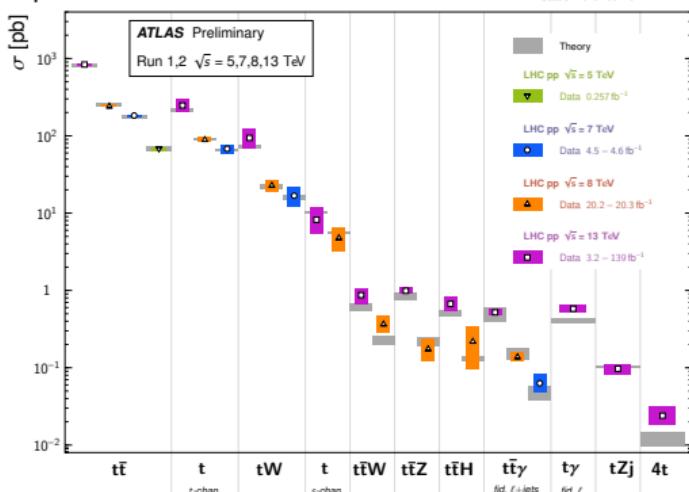
- Heaviest SM particle ( $172.67 \pm 0.30$  GeV)  
 ⇒ Almost unitary coupling with Higgs boson
- Decay:  $t \rightarrow Wb$  (BR=100%) with  $\tau_t = 5 \times 10^{-25}$  s ⇒ No bound states



# Top quark production at the LHC

## Top Quark Production Cross Section Measurements

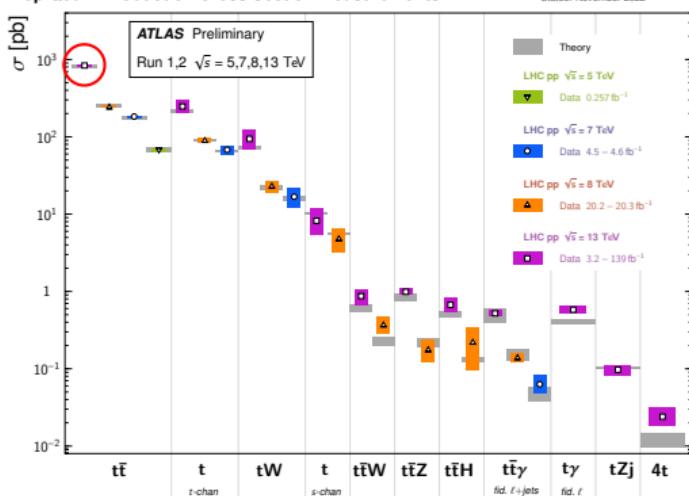
Status: November 2022



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## Production

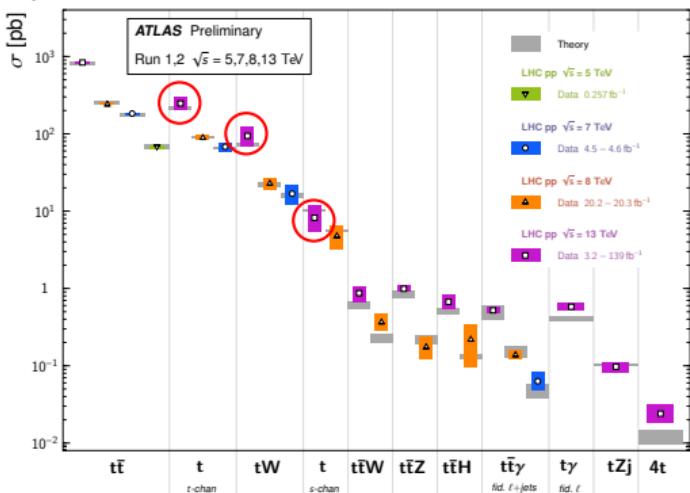
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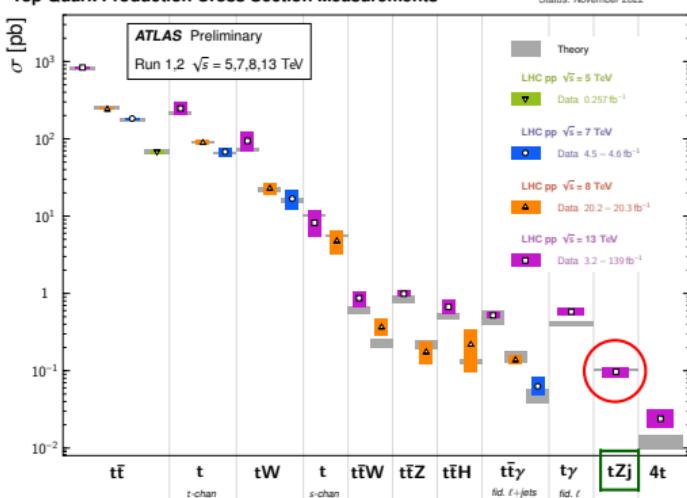


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  - $t$ -channel,  $tW$  and  $s$ -channel

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- Electroweak single top production:
  - $t$ -channel,  $tW$  and  $s$ -channel
- Associated productions: rarer processes, first observed at LHC thanks to unprecedent amount of data collected
  - $t\bar{t}W$ ,  $t\bar{t}Z$  and  $t\bar{t}H$
  - $tZq$



 Theory ②  $tZq$  associated production Event selection MVA training Fit Ratio of *top* and *antitop* cross sections in  $tZq$  production Summary and conclusions Backup

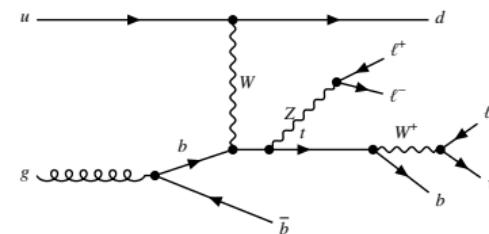
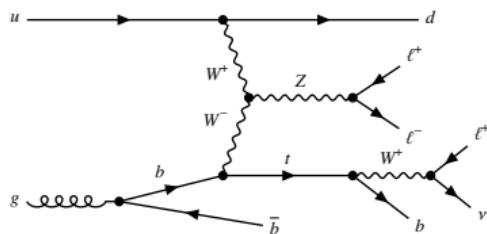
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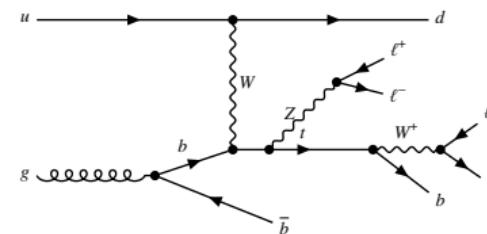
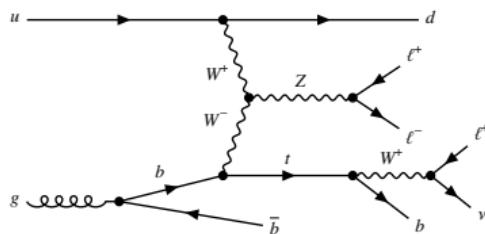
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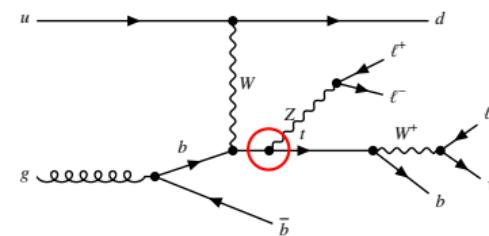
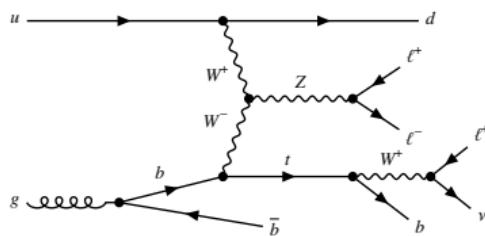
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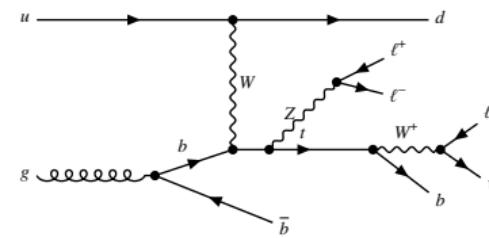
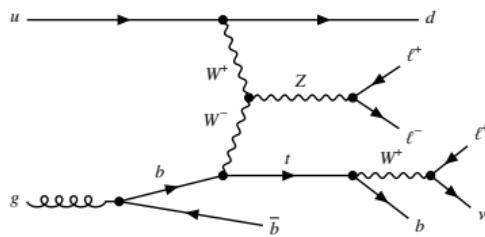
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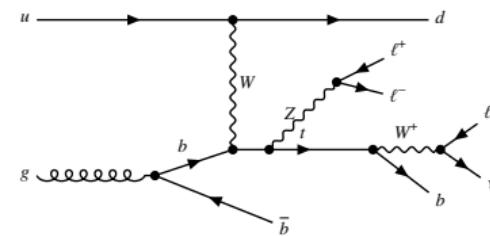
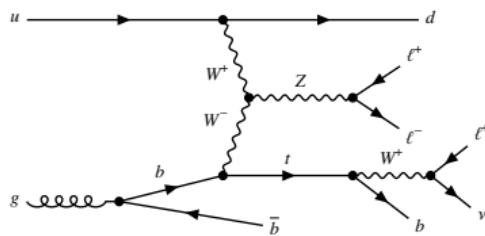
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- ATLAS measurement ( $139 \text{ fb}^{-1}$ ,  $\sqrt{s} = 13$  TeV):  
 $\sigma_{tZq} = 97^{+14}_{-13} \text{ fb} = 97^{+13}_{-12}(\text{stat}) \pm 7(\text{syst}) \text{ fb}$

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## Analysis strategy and goals

- Previous analysis measured the cross section with a 15% uncertainty



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  - Increasing signal acceptance by improving event selection
  - Better controlling of the backgrounds:
    - Increase separation between signal and backgrounds
    - Better modelling of specific sources

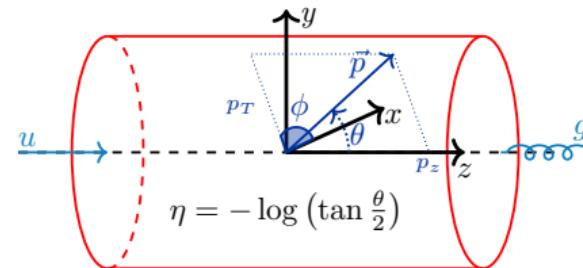
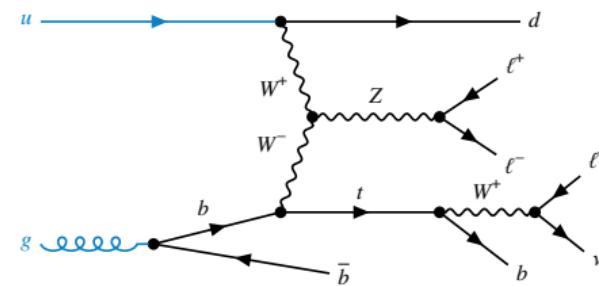


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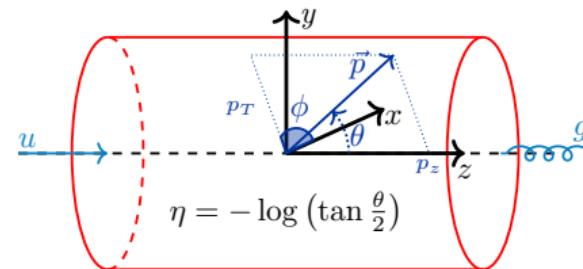
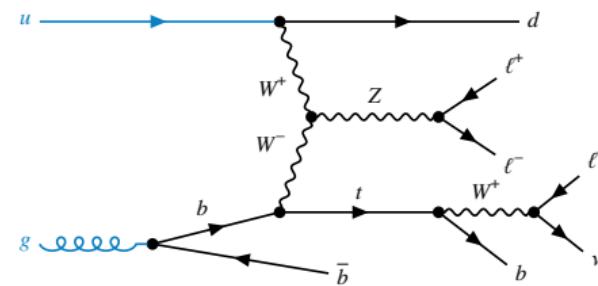
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- Exactly 3 leptons ( $e$  or  $\mu$ ):
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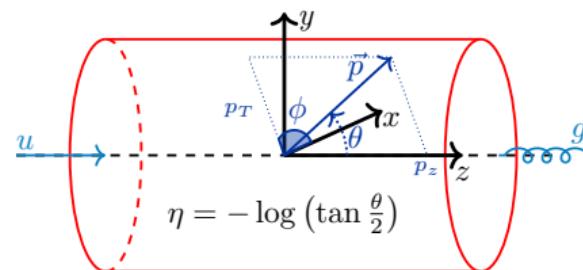
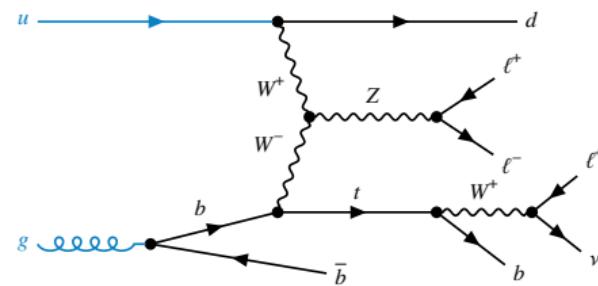
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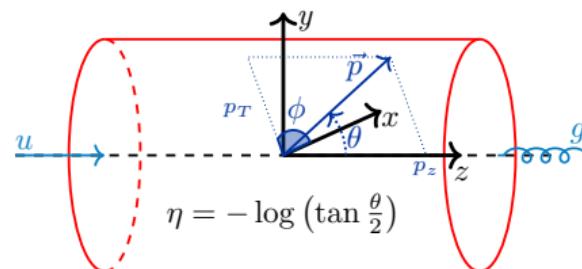
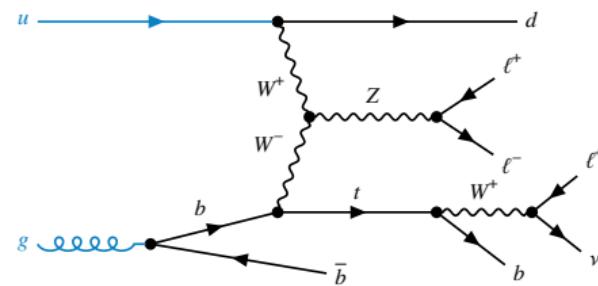
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- Multiple event selections tested, changing lepton  $p_T$  thresholds,  $b$ -jets efficiency and jet multiplicity



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Diboson plus jets

Fakes



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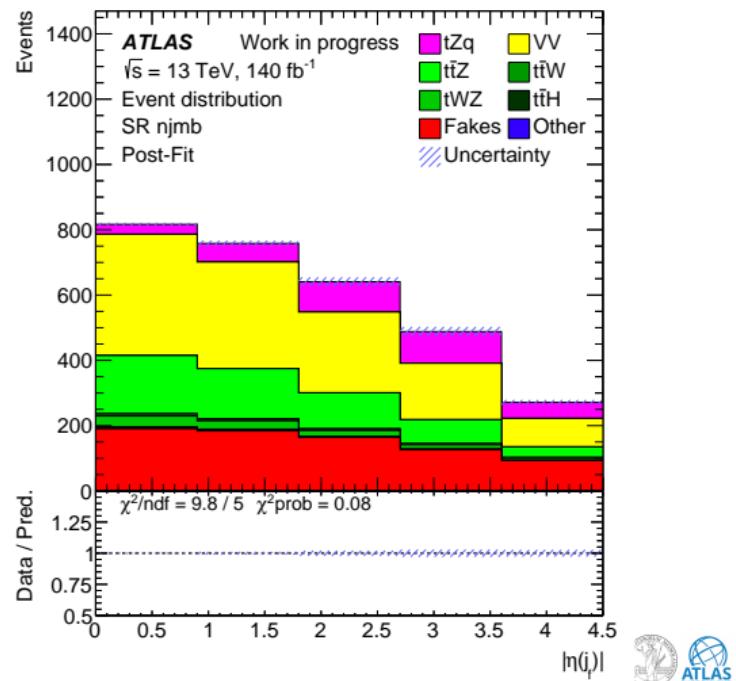
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- Other small backgrounds:  $tWZ$ ,  $t\bar{t}H$ ,  $t\bar{t}W$ , 3 and 4 tops production
- Our selection is loose on lepton  $p_T$  thresholds and  $b$ -jet identification efficiency, and also includes high jet multiplicity



# Event yields

Event yields		
$tZq$	$326.3 \pm 11.2$	
$t\bar{t}Z$	$549.5 \pm 26.3$	
Diboson	$1205.0 \pm 39.6$	
Fakes	$757.4 \pm 27.0$	
$t\bar{t}W$	$21.9 \pm 0.8$	
$tWZ$	$97.1 \pm 15.3$	
$t\bar{t}H$	$17.5 \pm 0.6$	
Other	$0.3 \pm 0.0$	
Total	$2974.9 \pm 100.5$	



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- Tested also a comparison with a Multi Class NN, which proved to be worse



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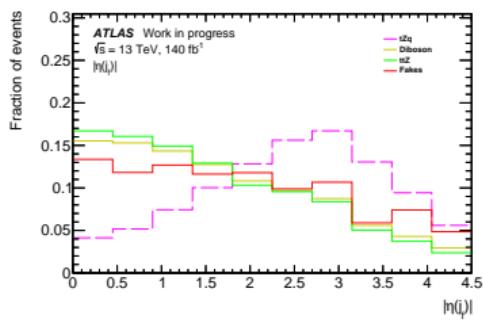
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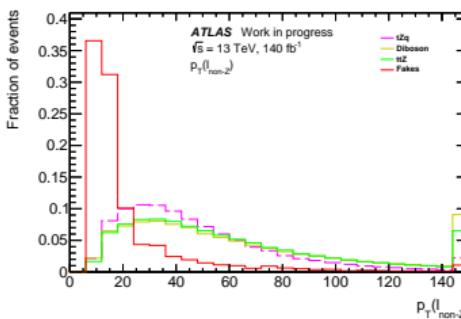
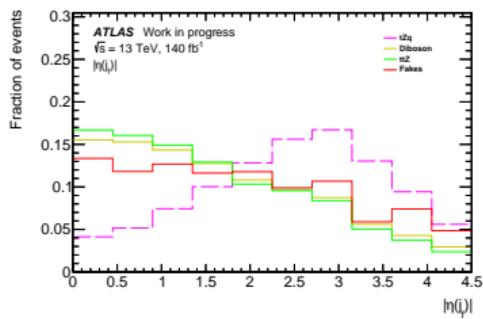
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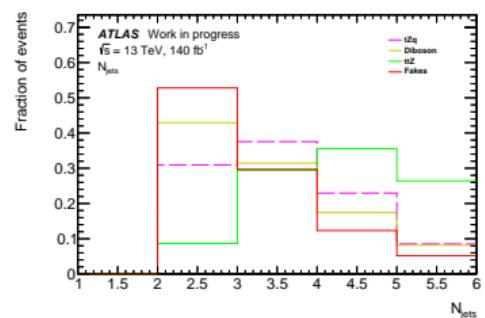
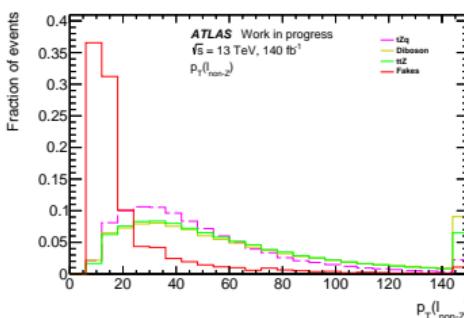
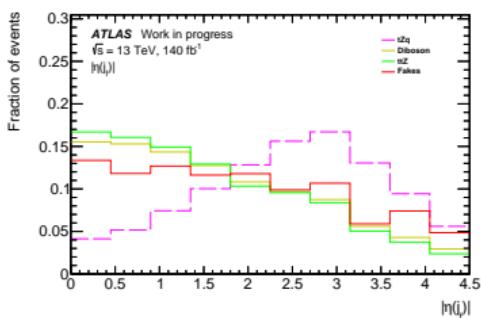
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  - $N_{\text{jets}}$ :  $t\bar{t}Z$  displays higher jet multiplicity

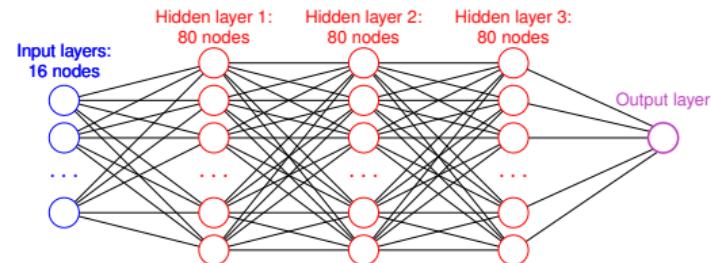


# Binary NN: setup and hyperparameters

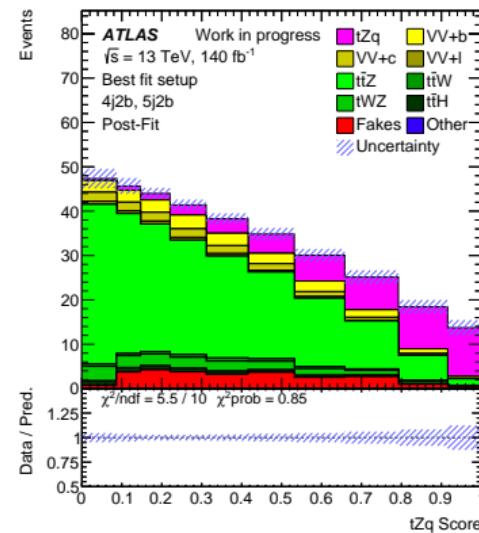
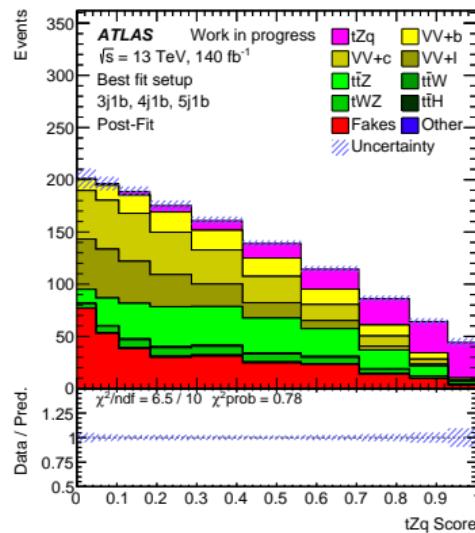
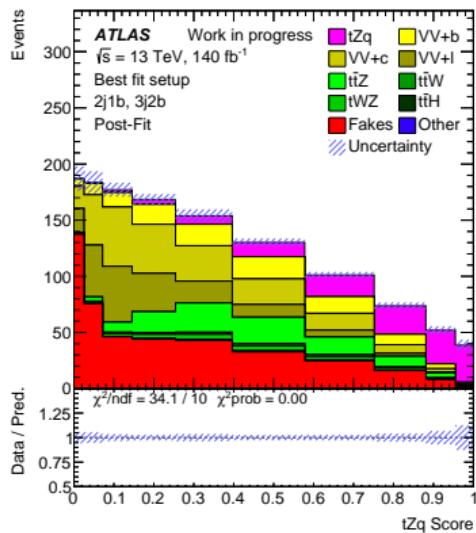
- Hyperparameters optimised using Grid search algorithm, tested 3600 configurations
  - Choice based on separation, AUC and loss function
- Used  $k$ -fold cross validation with  $k = 4$

## Hyperparameters

- 16 input variables
- 3 Hidden layers
- Nodes in each layer: 80,80,80
- Activation functions:  $eLu$ ,  $eLu$ ,  $eLu$
- Learning rate: 0.001
- Dropout rate: 0.2



# Binary training: splitting similar jet multiplicity events



NN output distributions in similar jet multiplicity events



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- Best values of the POI is the Maximum Likelihood Estimator (MLE), which is the parameter for which the Negative-Log-Likelihood reaches a minima
- Asimov fit  $\Rightarrow$  data substituted by Monte Carlo  $\Rightarrow$  POI MLE is 1



# Binned profile Likelihood fit

- Build a Likelihood which is the product of Poisson and Gaussian distributions
  - Poisson: signal plus background events in each bin
  - Gaussian: Nuisance Parameters ( $NP$ ,  $\theta_j$ ) distribution, referring to systematics uncertainties

$$\mathcal{L}(\vec{n}|\vec{\theta}, \vec{\mu}) = \prod_{i=1}^{N^{bins}} P\left(n_i|S_i(\vec{\theta}, \vec{\mu}) + B_i(\vec{\theta}, \vec{\mu})\right) \times \prod_{j=1}^{N^{NP}} G(\theta_j)$$

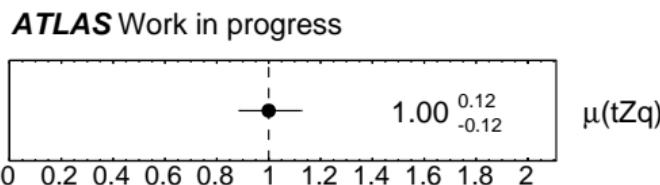
- Cross section extracted by a *binned profile Likelihood* fit, in which the Parameter of Interest (POI) is the signal strength  $\mu(tZq) = \frac{\sigma^{\text{measured}}}{\sigma^{\text{theory}}}$
- Best values of the POI is the Maximum Likelihood Estimator (MLE), which is the parameter for which the Negative-Log-Likelihood reaches a minima
- Asimov fit  $\Rightarrow$  data substituted by Monte Carlo  $\Rightarrow$  POI MLE is 1

Not presented

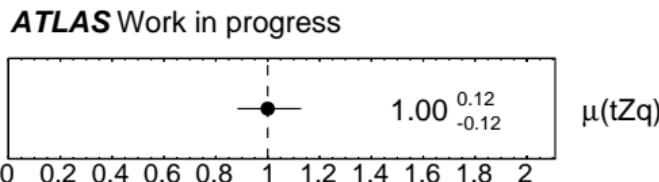
- Data-driven estimation of Diboson and Fakes to account for data/MC discrepancies



# Results



# Results



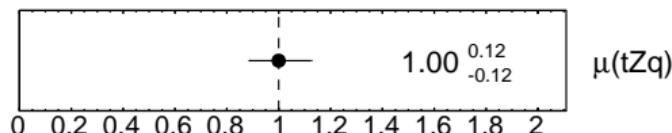
## Results

- MLE for the POI:  $\mu(tZq) = 1.00 \pm 0.12 = 1.00^{+0.10}_{-0.09}(\text{stat})^{+0.08}_{-0.07}(\text{syst})$



# Results

**ATLAS Work in progress**

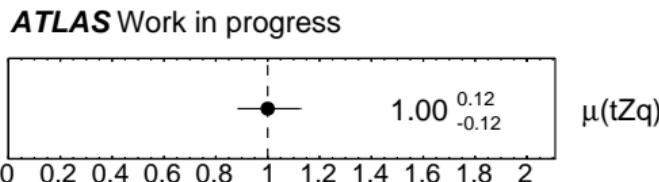


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  - Theoretical uncertainties dominate the systematics (6%, instrumental only 4%)



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  - Measurement is stat-dominated, just like previous analysis
  - Theoretical uncertainties dominate the systematics (6%, instrumental only 4%)
- 20% expected improvement with respect to the previous analysis (15%), using same dataset



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# Ratio of *top* and *antitop* cross sections in $tZq$ production

## Motivation

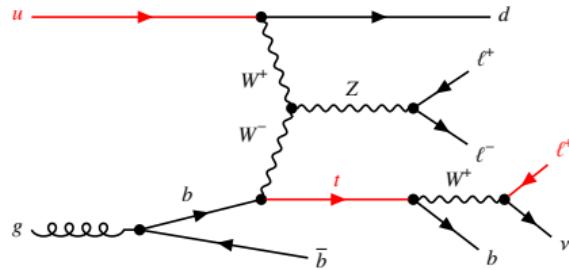
- $tZq$  process can occur with two different initial state quarks:



# Ratio of top and antitop cross sections in $tZq$ production

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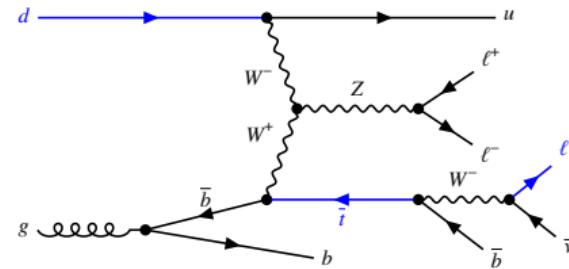
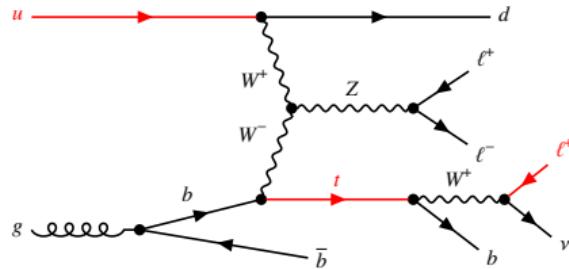
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  - “up” type (*u*, *c*)  $\Rightarrow$  *top* (*t*)  $\Rightarrow \ell^+$



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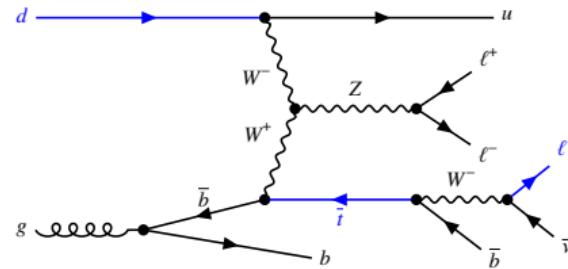
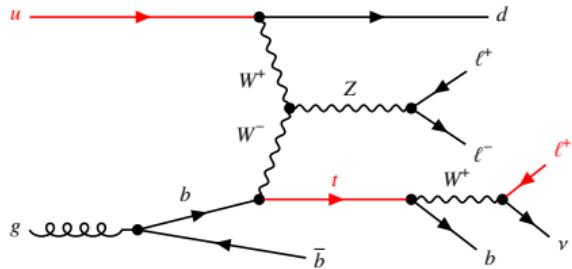
- $tZq$  process can occur with two different initial state quarks:
  - “up” type ( $u, c$ )  $\Rightarrow$  *top* ( $t$ )  $\Rightarrow \ell^+$
  - “down” type ( $d, s$ )  $\Rightarrow$  *antitop* ( $\bar{t}$ )  $\Rightarrow \ell^-$



# Ratio of top and antitop cross sections in $tZq$ production

## Motivation

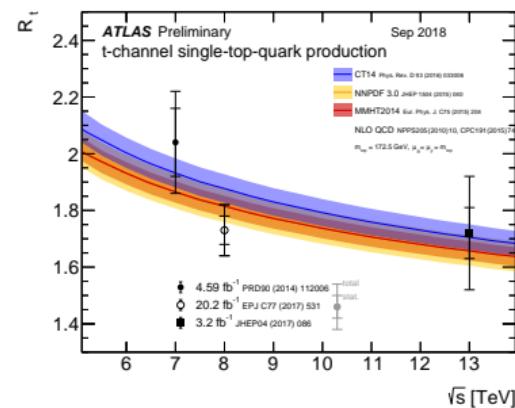
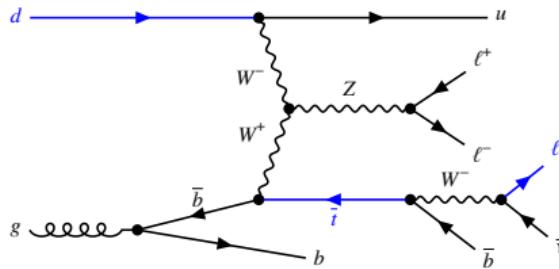
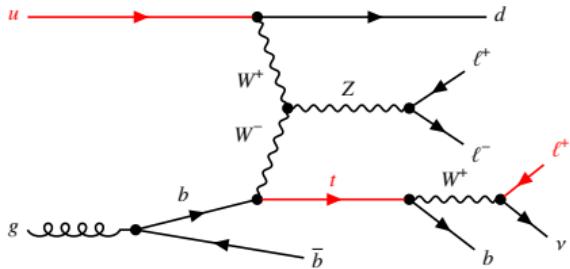
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- In proton, *up* more present than *down*  $\Rightarrow$  expected asymmetry, with the size of the asymmetry depending on proton internal structure



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- In proton, *up* more present than *down*  $\Rightarrow$  expected asymmetry, with the size of the asymmetry depending on proton internal structure
- Asymmetry measured in other channels but never in  $tZq$



# Ratio of *top* and *antitop* cross sections in $tZq$ production - results

## MLE for the POIs

- $\mu(tZq, t) = 1.00^{+0.14}_{-0.13} = 1.00 \pm 0.11(\text{stat}) \pm 0.07(\text{syst})$
- $\mu(tZq, \bar{t}) = 1.00^{+0.20}_{-0.18} = 1.00^{+0.17}_{-0.16}(\text{stat}) \pm 0.09(\text{syst})$
- $R(t/\bar{t}) = 1.00^{+0.23}_{-0.18} = 1.00^{+0.22}_{-0.17}(\text{stat}) \pm 0.06(\text{syst})$



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## Results

- These three measurements are the most precise in the  $tZq$  channel (uncertainties lower than CMS)



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## Future implications

- This selection and this MVA are currently being used in order to public ATLAS first  $tZq$  differential cross section measurement



Thank you for your attention  
Alberto Plebani



1 Theory

2  $tZq$  associated production

3 Event selection

4 MVA training

5 Fit

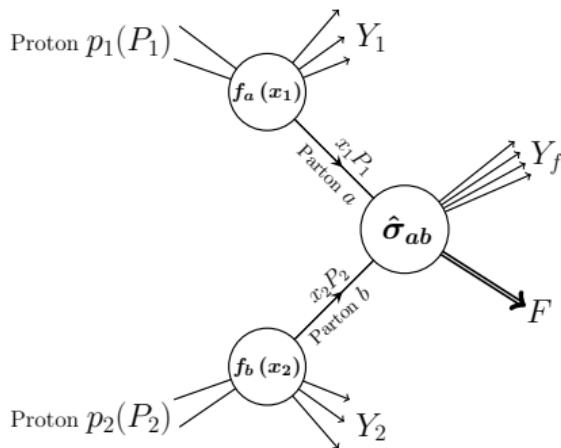
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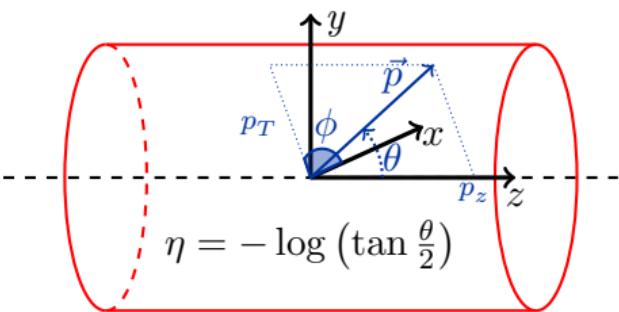
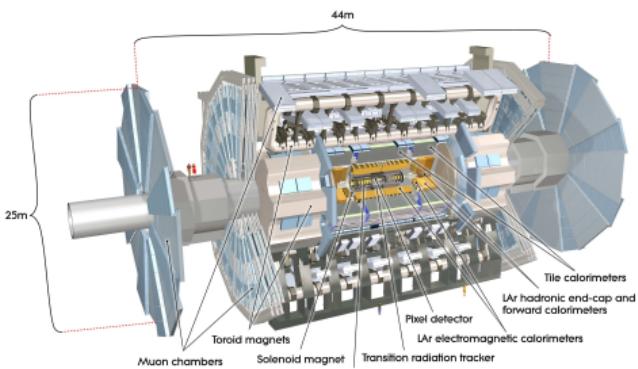
# Proton-proton inelastic cross section



$$\sigma(P_1, P_2; Q^2) = \sum_{a,b=q,\bar{q},g} \int_0^1 dx_1 dx_2 f_{a/p_1}(x_1, \mu_F^2) f_{b/p_2}(x_2, \mu_F^2) \hat{\sigma}_{ab}(x_1 P_1, x_2 P_2; \alpha_S(\mu_F^2), \mu_F^2, \mu_R^2, Q^2)$$



# The ATLAS experiment



## Multipurpose detector

- Higgs discovery, SM precision measurements
- BSM (SUSY, Dark Matter) searches

## Systems

- Magnetic: Toroidal and solenoidal
- Inner detector (ID)
- Electromagnetic Calorimeter (ECal)
- Hadronic Calorimeter (HadCal)
- Muon Spectrometer (MS)

## Data collected in Run2

- $140 \text{ fb}^{-1}$  of proton-proton inelastic collisions
- Centre-of-mass energy:  $\sqrt{s} = 13 \text{ TeV}$



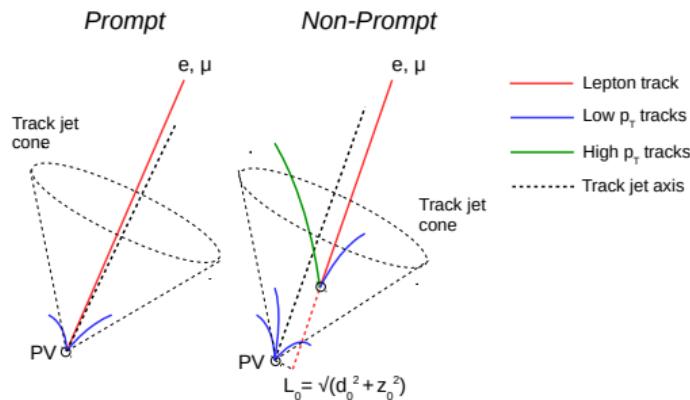
# Final States Branching Ratios

- 3 charged leptons ( $Z \rightarrow \ell^+ \ell^-$ ,  $W \rightarrow \ell \nu_\ell$ ):
  - Without hadronic  $\tau$  leptons: 1.8%
  - With hadronic  $\tau$  leptons: 3.2%
- 2 charged leptons ( $Z \rightarrow \ell^+ \ell^-$ ,  $W \rightarrow q\bar{q}'$ ):
  - Without hadronic  $\tau$  leptons: 4.6%
  - With hadronic  $\tau$  leptons: 6.8%
- 1 charged lepton ( $Z \rightarrow q\bar{q}$ ,  $W \rightarrow \ell \nu_\ell$ ):
  - Without hadronic  $\tau$  leptons: 22.8%
  - With hadronic  $\tau$  leptons: 16.9%
- 0 charged leptons ( $Z \rightarrow q\bar{q}$  or  $Z \rightarrow \nu_\ell \bar{\nu}_\ell$ ,  $W \rightarrow q\bar{q}'$ ):
  - Inclusive: 60.2%
  - Only  $Z$  visible: 44.8%
  - Only  $Z$  invisible: 13.4%



# Prompt vs non-prompt leptons

- Prompt leptons are the ones generated from  $W$  or  $Z$  boson decay
- Non-prompt leptons are instead generated inside a hadronic jet
- Therefore, prompt leptons are isolated, whereas non-prompt leptons are not



Comparison between prompt and non-prompt lepton



# Event Yields in different jet multiplicity regions

	SR 2j1b	SR 3j1b	SR 4j1b	SR 5j1b	SR 3j2b	SR 4j2b	SR 5j2b
$tZq$	$101.0 \pm 12.2$	$87.5 \pm 8.9$	$42.8 \pm 5.5$	$13.5 \pm 2.2$	$35.1 \pm 3.5$	$32.1 \pm 3.9$	$14.3 \pm 2.1$
$t\bar{t}Z$	$39.6 \pm 3.0$	$84.3 \pm 4.9$	$91.1 \pm 3.4$	$63.8 \pm 3.0$	$73.1 \pm 5.6$	$109.1 \pm 5.6$	$88.8 \pm 2.3$
Diboson	$517.3 \pm 7.0$	$347.5 \pm 4.9$	$185.3 \pm 2.7$	$83.6 \pm 1.8$	$32.0 \pm 0.5$	$24.6 \pm 0.5$	$14.9 \pm 0.2$
Fakes	$400.6 \pm 11.9$	$196.6 \pm 6.2$	$74.3 \pm 2.0$	$31.8 \pm 1.1$	$28.2 \pm 0.9$	$18.6 \pm 0.6$	$7.2 \pm 0.2$
$t\bar{t}W$	$5.2 \pm 0.1$	$3.8 \pm 0.1$	$2.0 \pm 0.0$	$1.1 \pm 0.0$	$5.1 \pm 0.1$	$3.0 \pm 0.1$	$1.6 \pm 0.0$
$tWZ$	$16.2 \pm 1.7$	$24.7 \pm 2.9$	$19.4 \pm 2.8$	$10.4 \pm 1.7$	$8.0 \pm 1.2$	$10.5 \pm 2.3$	$8.0 \pm 2.3$
$t\bar{t}H$	$1.6 \pm 0.0$	$2.6 \pm 0.1$	$2.6 \pm 0.1$	$1.8 \pm 0.0$	$2.8 \pm 0.1$	$3.3 \pm 0.1$	$2.8 \pm 0.1$
Other	$0.0 \pm 0.0$	$0.0 \pm 0.0$	$0.0 \pm 0.0$	$0.1 \pm 0.0$	$0.0 \pm 0.0$	$0.1 \pm 0.0$	$0.1 \pm 0.0$
Total	$1081.7 \pm 23.3$	$746.9 \pm 15.7$	$417.4 \pm 9.4$	$206.0 \pm 5.9$	$184.3 \pm 6.2$	$201.3 \pm 6.4$	$137.6 \pm 3.6$

- Other is 3 and 4 top quark productions
- $njm_b$  notation refers to events with  $n$  jets, of which  $m$   $b$ -jets



# Binary vs Multi Class

## Binary

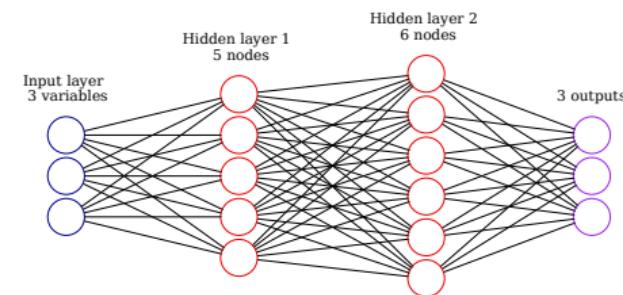
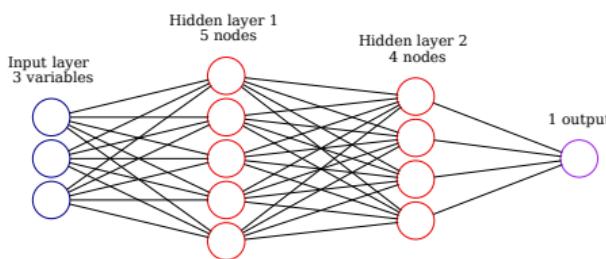
- 2 targets: signal trained versus all remaining backgrounds
- 1 output, with  $O \in [0, 1]$
- Greater separation
- Fit made using one region, in which the output was fitted
- Training only similar events improved the fit

## Multi Class (MC)

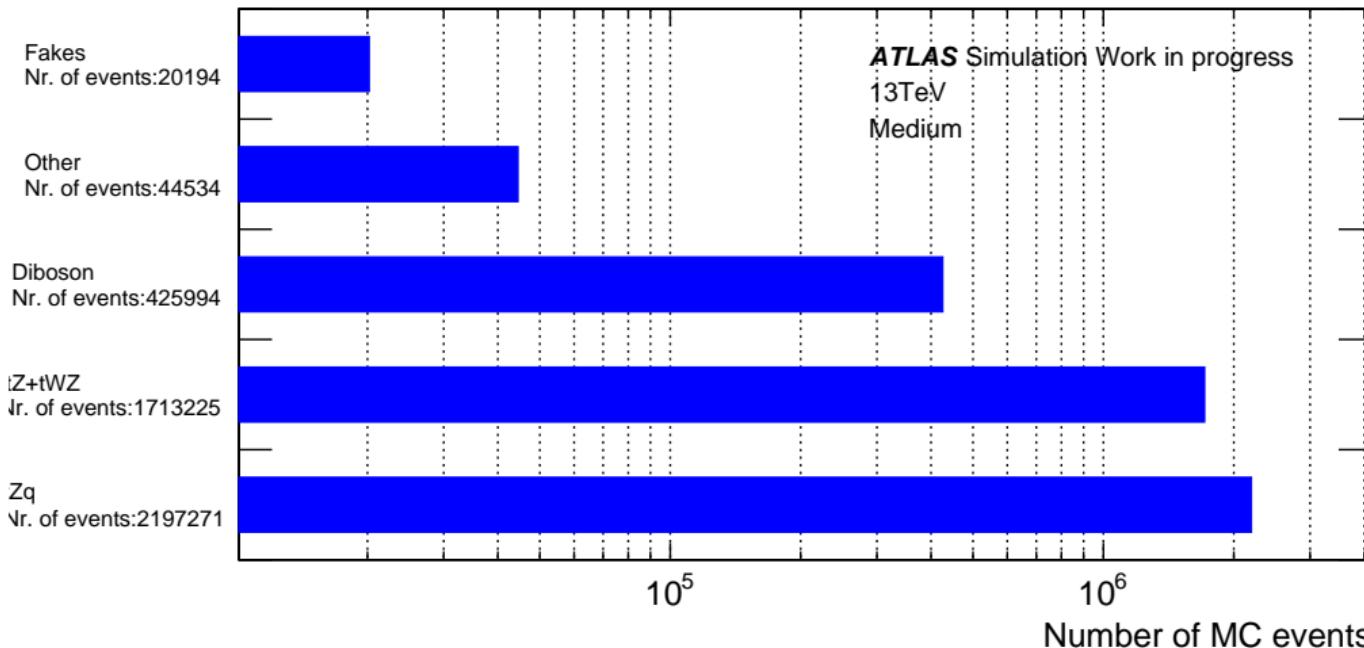
- 3 targets: signal,  $t\bar{t}Z$  or Diboson, all remaining backgrounds
- 3 outputs ( $O_i \leq 1$ ), with  $\sum O_i = 1$  for each event
- Lower separation
- Useful for the fits: allow to define different regions in order to maximise signal significance, and have background-enhanced regions
- Training only similar events improved the fit



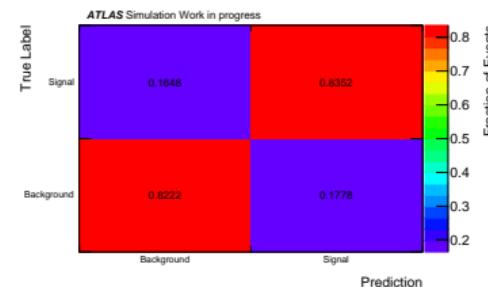
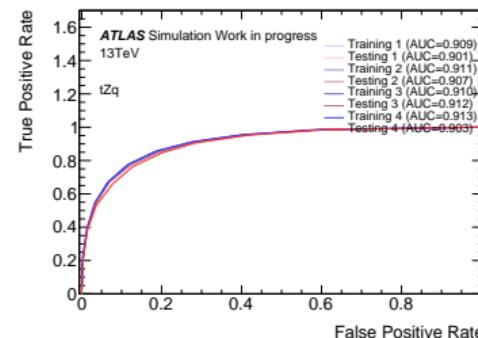
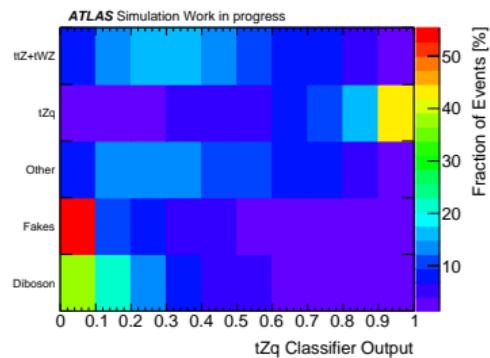
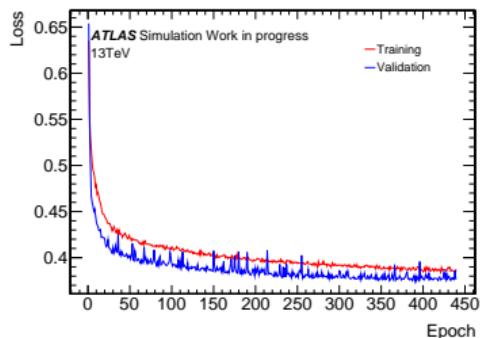
# Binary vs Multi Class



# Training dataset size



# NN specifics

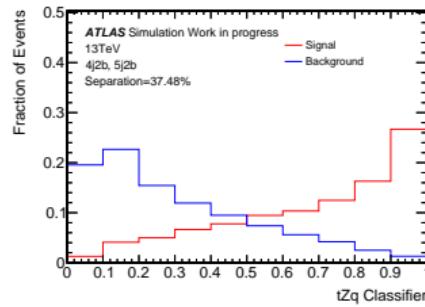
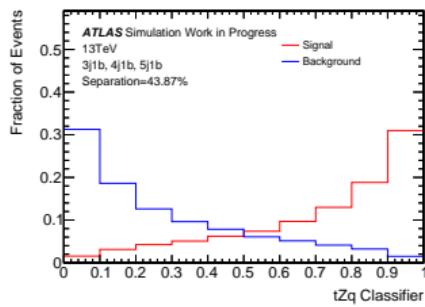
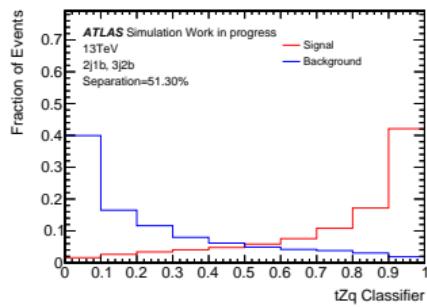


# Binary training: splitting similar jet multiplicity events

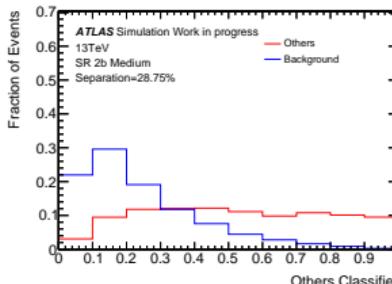
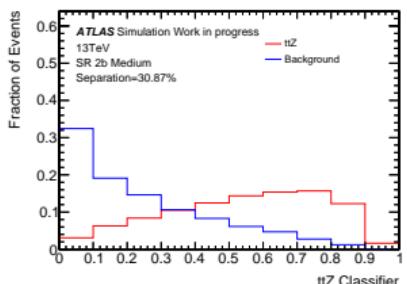
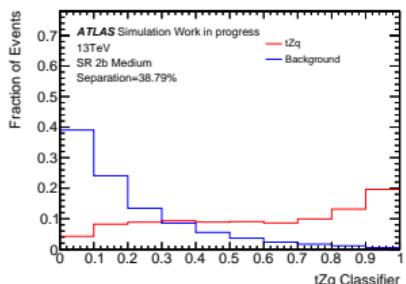
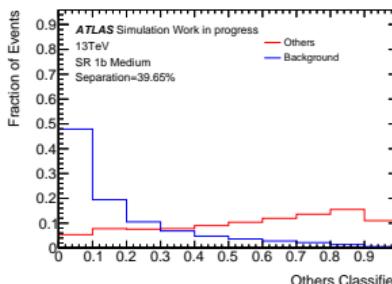
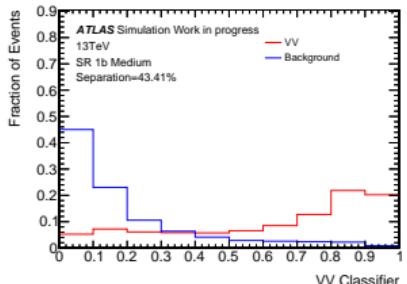
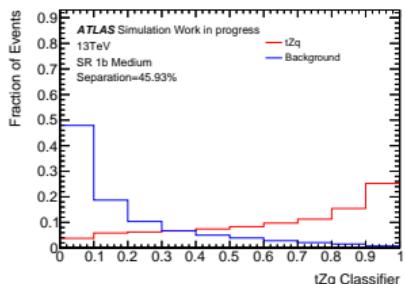
## Best training

- Three Binary NNs, each trained on similar jet multiplicity events:  
 $2j1b$  and  $3j2b$ ;  $3j1b$ ,  $4j1b$ , and  $5j1b$ ;  $4j2b$  and  $5j2b$
- We can also use radiation jet information in the latter two trainings (not present in  $2j1b$  and  $3j2b$ )

	$2j1b, 3j2b$	$3j1b, 4j1b, 5j1b$	$4j2b, 5j2b$
$tZq$	$136.2 \pm 29.8$	$143.7 \pm 21.2$	$46.4 \pm 10.9$
$VV + b$	$115.3 \pm 4.1$	$131.7 \pm 4.4$	$21.9 \pm 0.8$
$VV + c$	$242.1 \pm 8.4$	$268.8 \pm 8.8$	$13.4 \pm 0.5$
$VV + l$	$191.9 \pm 6.6$	$215.8 \pm 7.4$	$4.1 \pm 0.2$
$t\bar{t}Z$	$112.7 \pm 10.5$	$239.2 \pm 10.6$	$197.9 \pm 10.0$
$t\bar{t}W$	$10.3 \pm 0.4$	$7.0 \pm 0.3$	$4.6 \pm 0.2$
$tWZ$	$24.2 \pm 3.0$	$54.4 \pm 7.8$	$18.5 \pm 4.9$
$t\bar{t}H$	$4.4 \pm 0.2$	$7.0 \pm 0.2$	$6.1 \pm 0.2$
Fakes	$428.9 \pm 29.1$	$302.7 \pm 18.7$	$25.7 \pm 2.4$
Other	$0.0 \pm 0.0$	$0.1 \pm 0.0$	$0.2 \pm 0.0$
Total	$1266.0 \pm 58.4$	$1370.3 \pm 52.3$	$338.9 \pm 18.2$



# Multi Class training: splitting different $b$ -jet multiplicity events

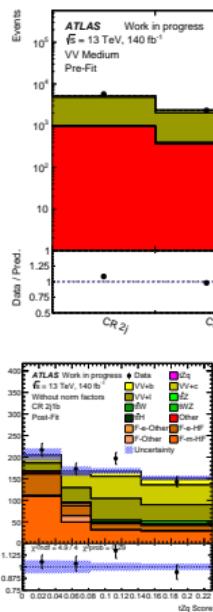


# Diboson background reweighting

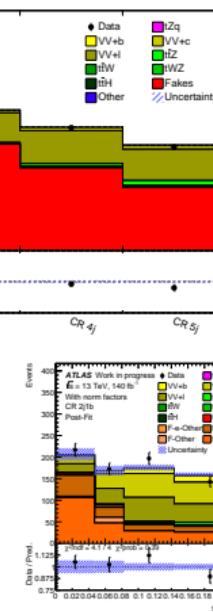
## Strategy

- Extract normalisation factors (NF) for Diboson using 0- $b$ -jets events (orthogonal selection wrt event selection)
- One NF for number of jets
- Validation: applying the NF to the Diboson MC simulated events to see if the Data/MC agreement improved

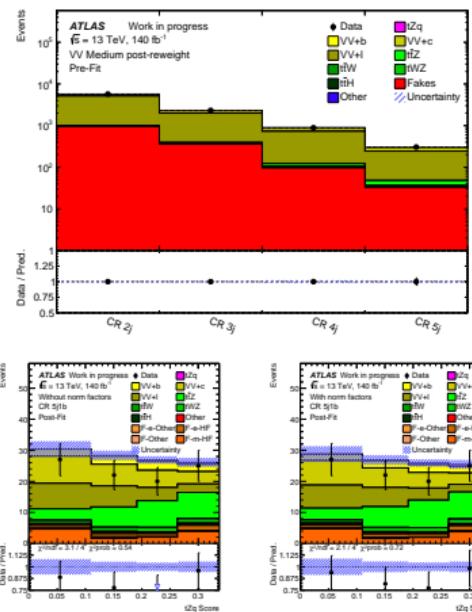
	2j	3j	4j	5j
NF	1.10	0.97	0.95	0.88



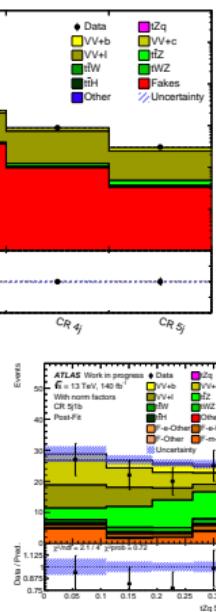
Before applying NFs



After applying NFs



Before applying NFs

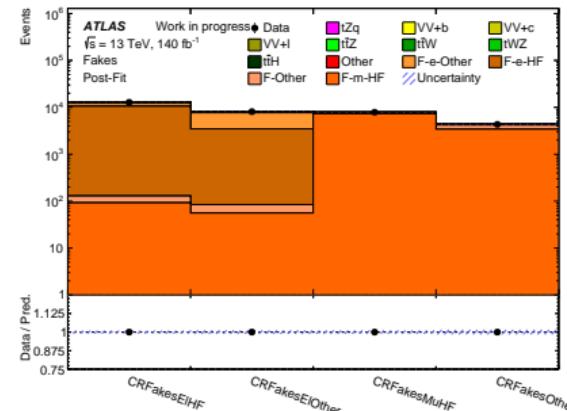
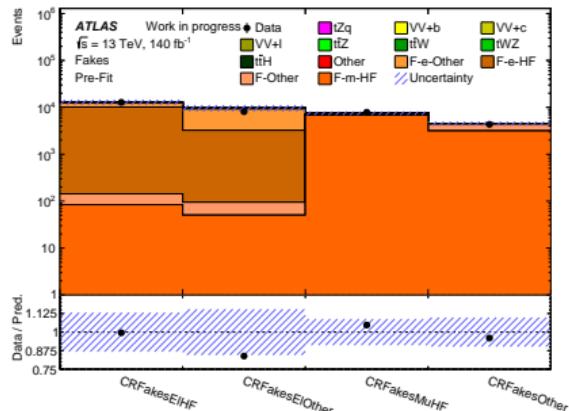


After applying NFs



# Fake background reweighting

Regions definition using <i>Medium</i> selection				
SR	<i>CR Fakes El HF</i>	<i>CR Fakes El Other</i>	<i>CR Fakes <math>\mu</math> HF</i>	<i>CR Fakes Other</i>
3 tight leptons ( $e$ or $\mu$ )	Exactly 2 tight leptons ( $e$ or $\mu$ )		Loose non tight lepton is electron	
2-5 jets	Loose non tight lepton is muon		2-5 jets	
1-2 $b$ -jets (2 $j2b$ excluded)	$\geq 1$ $b$ -jets	0 $b$ -jets	$\geq 1$ $b$ -jets	0 $b$ -jets
$ m_{\ell\ell} - m_Z  < 10$ GeV	$ m_{\ell\ell} - m_Z  > 10$ GeV	$ m_{\ell\ell} - m_Z  < 10$ GeV $m_T(W) < 20$ GeV	$ m_{\ell\ell} - m_Z  > 10$ GeV	$ m_{\ell\ell} - m_Z  < 10$ GeV $m_T(W) < 20$ GeV
-	-	Fit on number of jets	Fit on $p_T(\ell_3)$	Fit on number of jets
-	-	Fit on $p_T(\ell_3)$	Fit on number of jets	Fit on $p_T(\ell_3)$



# List of uncertainties

## Theoretical uncertainties

- Parton Distribution Functions (PDF): uncertainty on proton's internal structure
- Scale variation: uncertainty on the QCD coupling constant, depending on renormalisation and factorisation scales
- Showering: uncertainty on the parton shower generated after the collision, sensitive to jet multiplicity
- Initial state radiation: uncertainty on QCD radiation prior to the collision

## Instrumental uncertainties

- Luminosity: uncertainty on number of total events
- Pile-up reweighting: uncertainty on additional interactions occurring in the same time window of the main collision
- Jets: uncertainty on hadronic jets identification and reconstruction
- $b$ -tagging: uncertainty on the jets flavour tagging
- Leptons: uncertainties on the identification, isolation and reconstruction of electrons, muons, and photons

Group	Sub-group	Nuisance Parameters		Group	Sub-group	Nuisance Parameters
Theory (28)	$tZq$	4 (PDF, PS, ISR, scale)	Instrumental (158)	Instrumental (158)	Leptons	23
	$t\bar{t}Z$	4 (PDF, PS, ISR, scale)			Jets	48
	$tWZ$	2 (Normalisation, diagram removal)			$b$ -tagging	85
	$VV$	6 (PDF, scale; 2 for each flavour)			Pile-Up modelling	1
	$t\bar{t}$	12 (PS, Generator, ISR; 3 for each flavour)			Luminosity	1

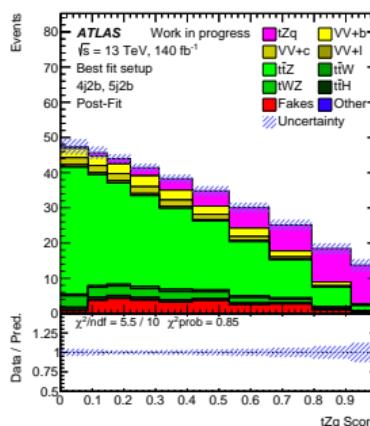
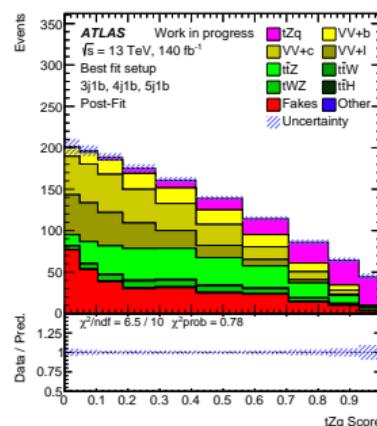
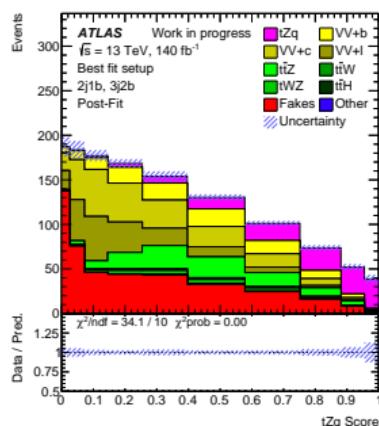
Total = 186 Nuisance Parameters for systematics uncertainties



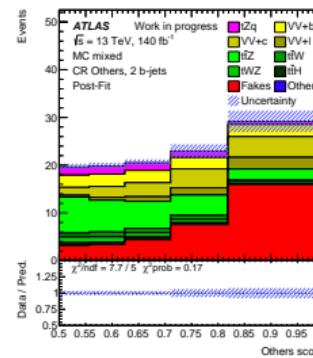
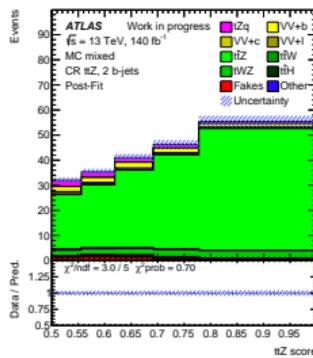
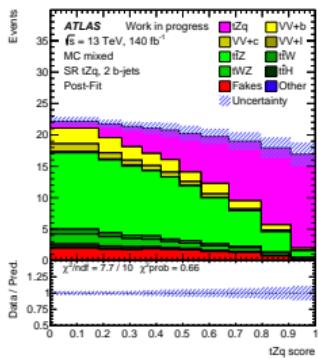
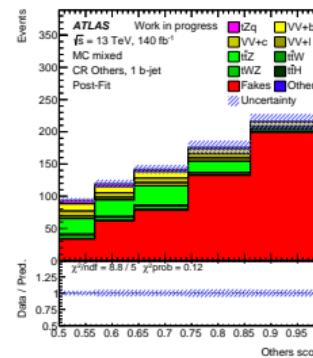
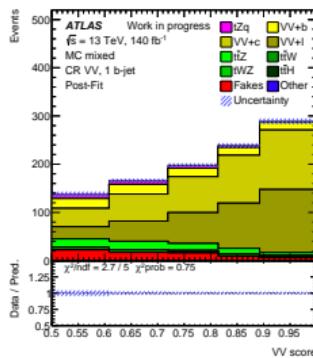
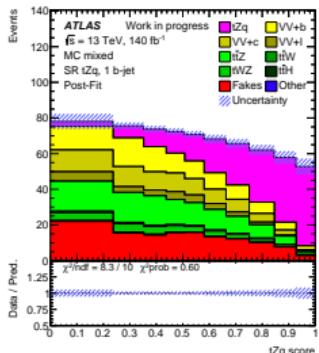
# Fitting Binary NN trained on similar jet multiplicity events

## Fit setup

- Asimov fit is blind  $\Rightarrow$  data not used for the fit, and not visible on plots
- Signal Regions (SR) defined by grouping events the same way they were for the training (page 15):  $2j1b$  and  $3j2b$ ;  $3j1b$ ,  $4j1b$ , and  $5j1b$ ;  $4j2b$  and  $5j2b$
- In each group the fit was made to the output of the Binary NN trained on those events



# Fit using Multi Class training, splitting different $b$ -jet multiplicity events



# Ratio of top and antitop cross sections in $tZq$ production - 2

## Methods

- As aforementioned, the difference between top and antitop is the charge of  $\ell_W$
- We split the dataset based on that charge, and obtained the following values
- We observe the asymmetry in simulated signal events, with a ratio of 1.8
- Diboson asymmetric due to  $WZ$
- $t\bar{t}Z$  and Fakes symmetric because generated by gluon-gluon fusion, and thus they do not depend on the PDFs

	$t, Q(\ell_W) = +1$	$\bar{t}, Q(\ell_W) = -1$	Ratio $t/\bar{t}$
$tZq$	$207.9 \pm 20.8$	$118.4 \pm 11.8$	1.8
$VV + b$	$156.7 \pm 3.2$	$112.2 \pm 2.3$	1.4
$VV + c$	$279.3 \pm 5.6$	$245.0 \pm 5.2$	1.1
$VV + l$	$244.5 \pm 5.4$	$167.3 \pm 3.9$	1.5
$t\bar{t}Z$	$281.5 \pm 10.8$	$268.1 \pm 9.4$	1.0
$t\bar{t}W$	$14.3 \pm 0.3$	$7.6 \pm 0.2$	2.0
$tWZ$	$48.6 \pm 7.4$	$48.5 \pm 7.1$	1.0
$t\bar{t}H$	$8.8 \pm 0.2$	$8.7 \pm 0.2$	1.0
Fakes	$375.5 \pm 10.2$	$381.8 \pm 10.2$	1.0
Other	$0.2 \pm 0.0$	$0.2 \pm 0.0$	1.0
Total	$1617.4 \pm 28.9$	$1357.8 \pm 24.4$	1.2



# Ratio of top and antitop cross sections in $tZq$ production - results

## MLE for the POIs

- $\mu(tZq, t) = 1.00^{+0.14}_{-0.13} = 1.00 \pm 0.11(\text{stat}) \pm 0.07(\text{syst})$
- $\mu(tZq, \bar{t}) = 1.00^{+0.20}_{-0.18} = 1.00^{+0.17}_{-0.16}(\text{stat}) \pm 0.09(\text{syst})$
- $R(t/\bar{t}) = 1.00^{+0.23}_{-0.18} = 1.00^{+0.22}_{-0.17}(\text{stat}) \pm 0.06(\text{syst})$

## Considerations

- Systematics uncertainties tend to cancel in the ratio  $\Rightarrow$  much dominated by statistics
- The three cross sections are dominated by theoretical uncertainties, particularly signal modelling uncertainty
- Top, Antitop and ratio uncertainties represent the most precise measurements in the  $tZq$  channel

Systematic	Inclusive	Top	Antitop	Ratio
$tZq$ modelling	5.0%	4.8%	5.7%	1.0%
Diboson+ $t\bar{t}Z$ modelling	3.3%	3.0%	4.2%	1.7%
Leptons	3.3%	3.2%	3.6%	0.5%
Jets	1.6%	1.6%	2.2%	1.6%
Fakes modelling	1.6%	2.5%	2.9%	4.6%
$b$ -tagging	1.3%	1.4%	1.4%	0.8%
MC statistics	1.0%	1.3%	2.4%	2.7%
Luminosity	0.8%	0.8 %	0.8%	0.1%
Pileup	0.8%	0.9%	0.8%	0.1%
$tZq$ PDF	0.3%	0.4%	0.6%	0.2%
Theory	6.1%	6.1%	7.5%	5.1%
Instrumental	4.1%	4.2%	5.0%	2.9%
Full systematics	7.4%	7.4%	9.0%	5.9%
Statistical	9.5%	11.0%	16.5%	19.5%
Total	12.0%	13.5%	19.0%	20.5%

