

# Measurements of $W^+W^-$ production cross sections at $\sqrt{s} = 13$ TeV with the ATLAS detector



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The  
University  
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Sheffield.

# Introduction

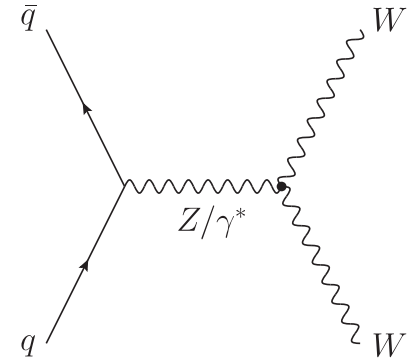
- ▶ **WW measurements provide precision tests of Standard Model (SM)**

- Sensitive to properties of gauge boson self-interactions
- Test of perturbative quantum chromodynamics (pQCD) and electroweak (EW) theory

- ▶ **Important background for  $H \rightarrow WW$  measurements and BSM searches**

- ▶ **Previous WW measurements at the LHC:**

- $\sqrt{s} = 7$  TeV    [arXiv:1210.2979](#)    [arXiv:1306.1126](#)    2013
- $\sqrt{s} = 8$  TeV    [arXiv:1603.01702](#)    [arXiv:1507.03268](#)    [arXiv:1608.03086](#)    ↓
- $\sqrt{s} = 13$  TeV    [arXiv:1702.04519](#)    [arXiv:1905.04242](#)    [arXiv:2009.00119](#)    2020
- All limit number of hadronic jets to reduce backgrounds



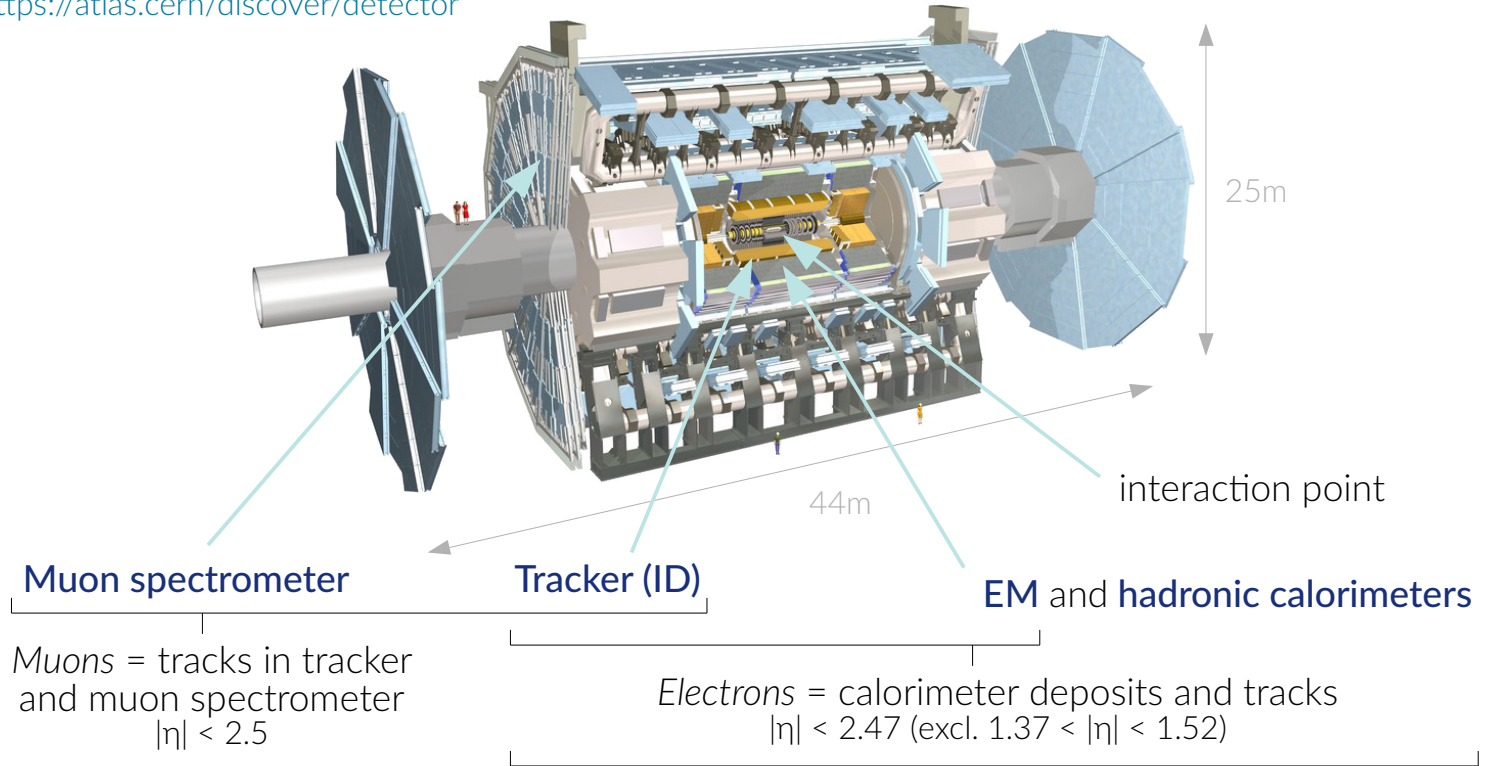
**Most recent ATLAS measurement inclusive over jets** → focus of this talk

2021

└─▶ shown at [Moriond](#) in March 2021

# ATLAS detector and reconstruction

<https://atlas.cern/discover/detector>



[arXiv:1703.10485](https://arxiv.org/abs/1703.10485)  
[arXiv:0802.1189](https://arxiv.org/abs/0802.1189)

← *Jets* = (anti- $k_T$ ) clustered calorimeter deposits and tracks  
 $|\eta| < 4.5$

[ATL-PHYS-PUB-2017-013](https://arxiv.org/abs/ATL-PHYS-PUB-2017-013)  
[arXiv:1907.05120](https://arxiv.org/abs/1907.05120)

← Secondary vertices from  $b$ -hadron decays allow for  $b$ -jet identification (DL1r)

## Motivation

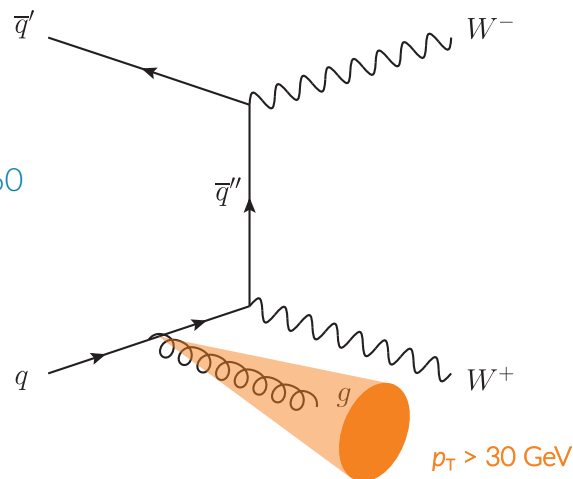
- ▶ **Jet-inclusive differential measurements made for first time at LHC**
- ▶ **Improved precision in fully inclusive measurement** (when combined with jet veto measurement)
- ▶ **Improved sensitivity to BSM physics**

- Effective field theory (EFT) interference term less helicity suppressed than in jet veto case

↳ [arXiv:1707.08060](https://arxiv.org/abs/1707.08060)

## Analysis strategy

- ▶ **Count  $pp \rightarrow e\nu\mu\nu$  (+ jets) events**
  - Data binned in 12 observables\*
- ▶ **Estimate backgrounds**
  - Dominant contribution from  $t\bar{t}$  events  $\rightarrow$  estimate with data-driven method
  - Fakes (data-driven), Z+jets, diboson,  $V\gamma$
- ▶ **Unfolded result = detector<sup>-1</sup> ( data - backgrounds )**



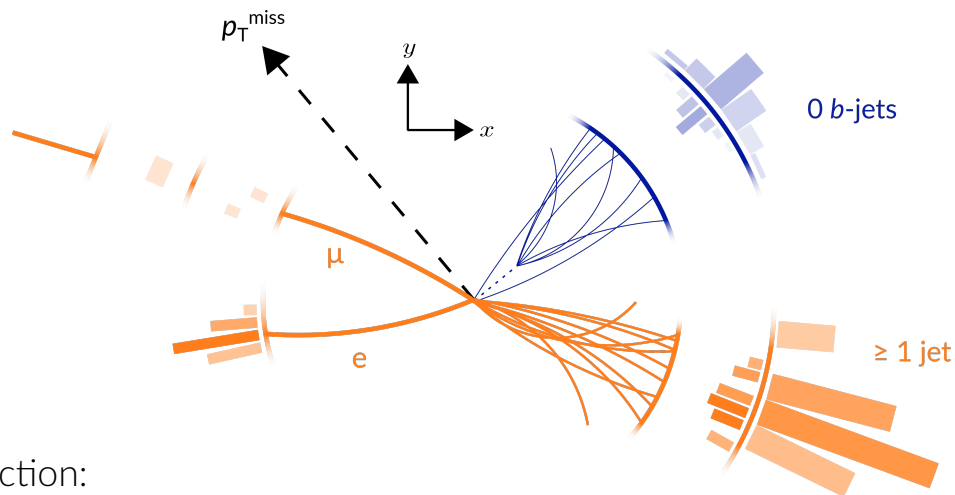
\*  $p_T^{\text{lead.lep.}}$ ,  $p_T^{\text{sublead.lep.}}$ ,  $p_T^{\text{lead.jet.}}$ ,  $n$  jets,  $m_{e\mu}$ ,  $p_{T,e\mu}$ ,  $H_T$ ,  $S_T$ ,  $m_{T,e\mu}$ ,  $\Delta\phi_{e\mu}$ ,  $\cos\theta^*$

# Event selection

- ▶ Lepton and jet cuts:

$$p_T^e, p_T^\mu > 27 \text{ GeV}$$

$$\geq 1 \text{ jet}$$
$$p_T > 30 \text{ GeV}$$



- ▶ Selections for background rejection:

opposite flavour leptons



Rejects:  $Z/\gamma^* \rightarrow ee/\mu\mu$

$$m_{e\mu} > 85 \text{ GeV}$$



Rejects:  $Z/\gamma^* \rightarrow \tau\bar{\tau}$  where  $\tau \rightarrow e/\mu (+ \nu\bar{\nu})$

b-jet veto

$$p_T > 20 \text{ GeV}$$



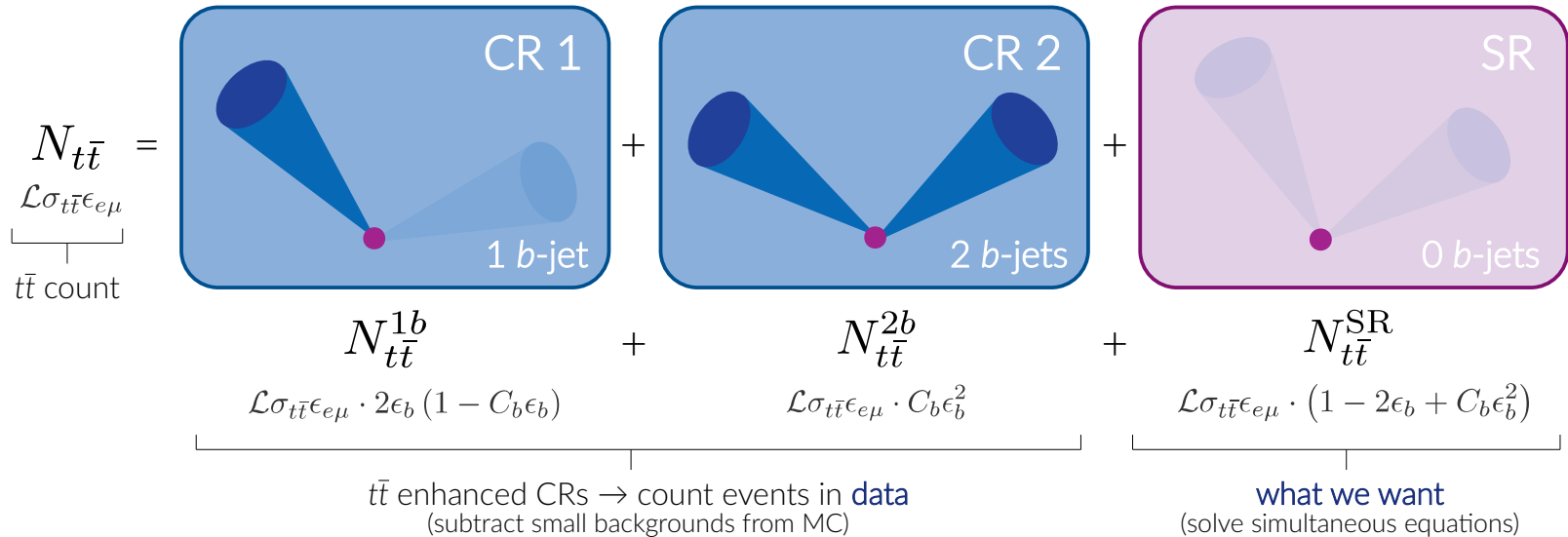
Rejects:  $t\bar{t}$  and  $Wt$

# Top estimate ( $t\bar{t} + Wt$ )

► **Top events account for ~61% of events in signal region (SR)**

→ [arXiv:1910.08819](https://arxiv.org/abs/1910.08819)

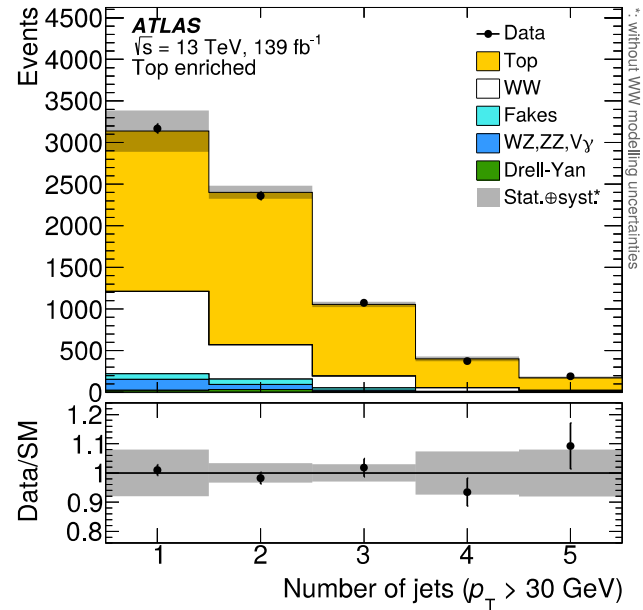
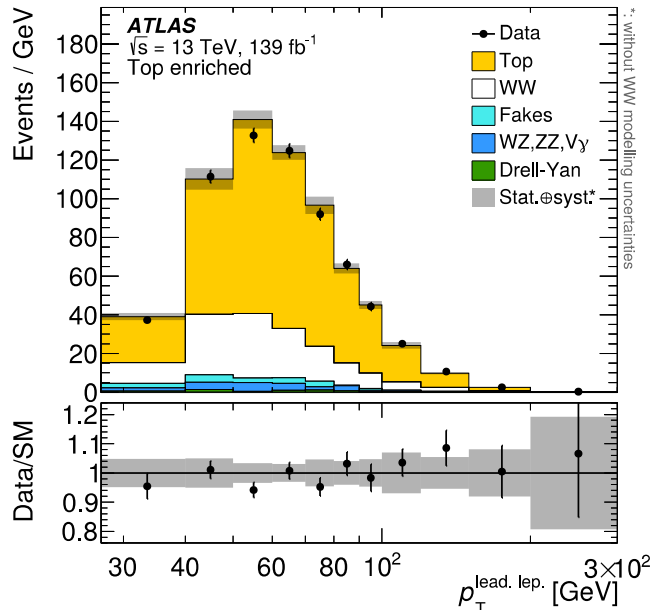
- Use data-driven ‘ $b$ -tag counting’ method inspired by  $t\bar{t}$  cross-section measurement
- Two control regions (CRs) with different numbers of *tagged*  $b$ -jets



- $t\bar{t}$  modelling only enters in  $C_b$
- **Single top ( $Wt$ ) contribution from MC**

$\epsilon_b = b$ -jet selection efficiency  
 $C_b =$  correlation factor (MC)

- ▶ **Estimate strongly reduces systematic uncertainties:** 15% (pure MC)  $\rightarrow$  2.8%
  - Anti-correlation between some  $t\bar{t}$  and  $Wt$  systematics reduces total uncertainty
- ▶ **Extensive closure tests performed**
- ▶ **Check estimate in top enriched validation region (VR):**  $m_{lj} < 140$  GeV,  $\Delta\phi_{e\mu} < \pi/2$  (+ SR)



# Fake estimate

- ▶ **Fake = jet misidentified as lepton / lepton from heavy flavour (HF) decay**, e.g.  $W$ +jets (3%)
- ▶ **Poorly modelled** → estimate contribution with data-driven fake factor (FF) method
  - Use two auxiliary regions: *dijet* and *ID+anti-ID*

**Dijet region** (dominated by two-jet events)

Extract **fake factors** here

$$FF = \frac{N_{ID} - N_{ID,MC}^{prompt}}{N_{anti-ID} - N_{anti-ID,MC}^{prompt}} = \frac{ID}{anti-ID}$$

**ID+anti-ID region** (SR with one ID → anti-ID)

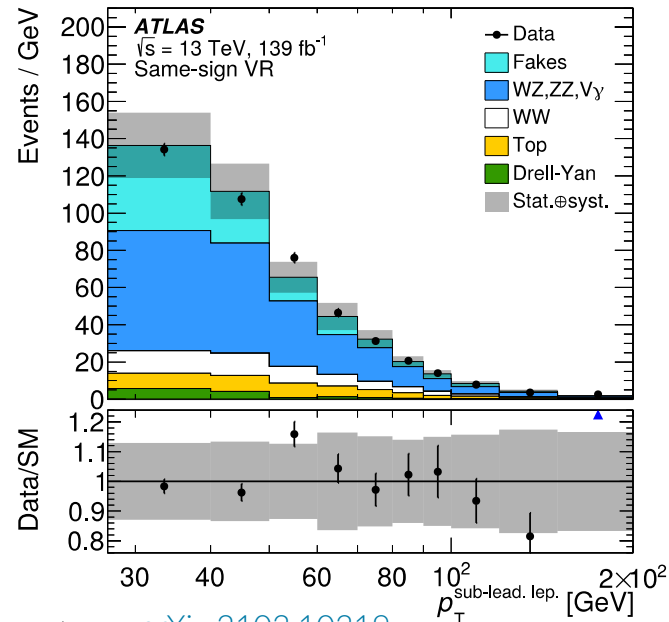
Apply fake factors here

$$N_{ID+ID}^{non-prompt} = FF \times (N_{ID+anti-ID} - N_{ID+anti-ID,MC}^{prompt})$$

$$= \frac{ID}{anti-ID} \times (ID+anti-ID) = (ID+ID)$$

- ▶ **FF estimate validated in same sign region**

$e/\mu$  selection in SR = ID  
orthogonal selection = anti-ID



fake count  
in SR

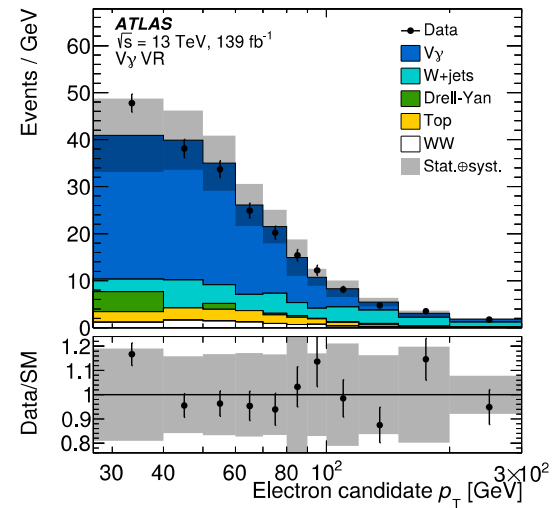
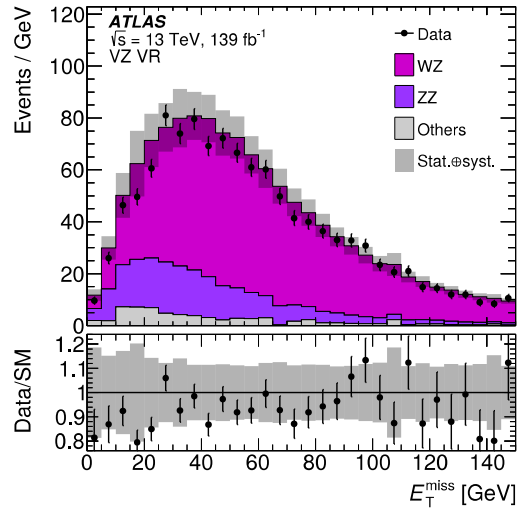
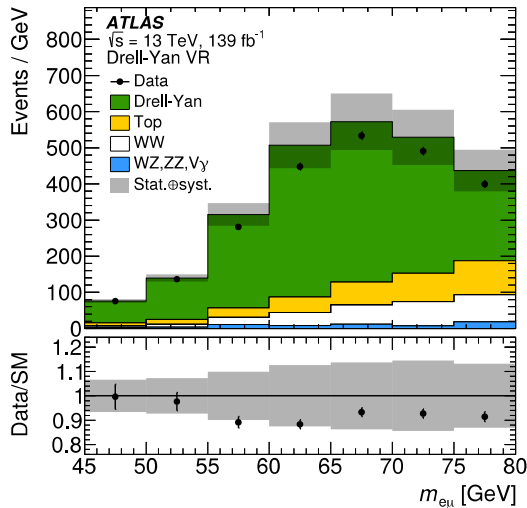
arXiv:2103.10319



# Other backgrounds

- ▶ All remaining backgrounds estimated from simulation and validated in dedicated VRs

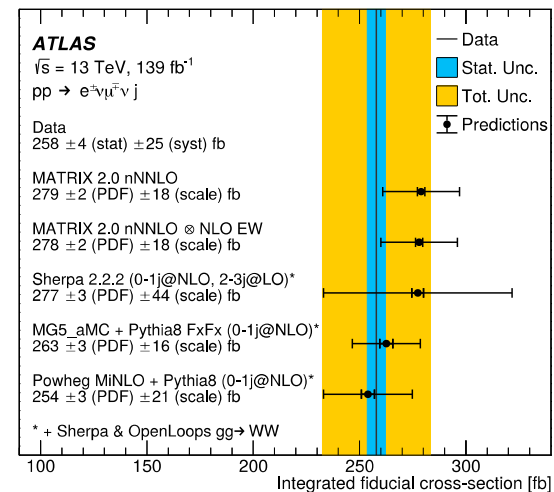
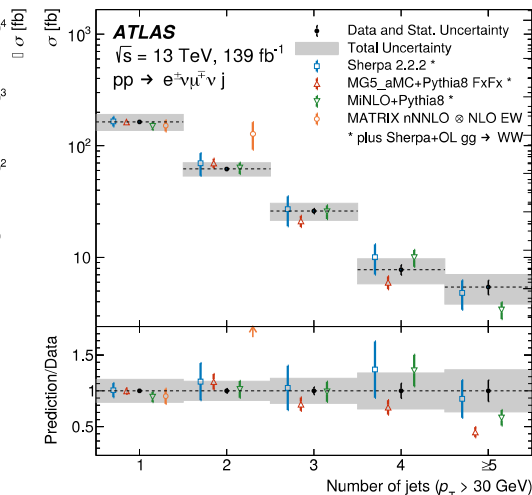
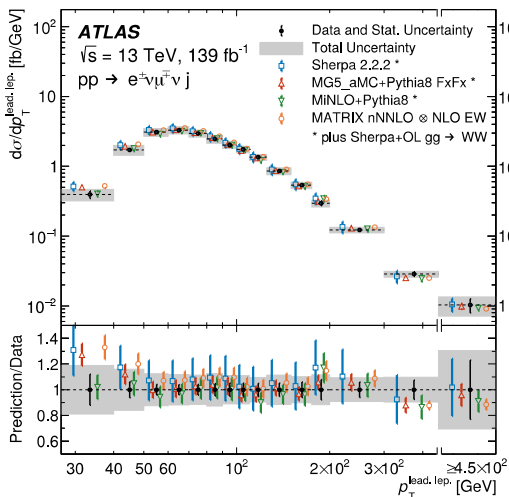
- Account for ~3% of events in SR
- Z+jets (Drell-Yan), VZ,  $V\gamma$
- Triboson negligible (< 0.1% of selected events)



[arXiv:2103.10319](https://arxiv.org/abs/2103.10319)

# Results

- ▶ **Correct for detector effects** using iterative Bayesian unfolding method → D'Agostini, 95 arXiv:1010.0632
- ▶ **Unfolded results obtained for fiducial and differential cross sections**
  - $\sigma_{\text{fid}} = 258 \pm 4$  (stat.)  $\pm 25$  (syst.) fb
  - Total uncertainty of  $\sim 10\%$  driven by jet calibration, top and fake contributions
- ▶ **Excellent agreement seen with theoretical predictions**



arXiv:2103.10319

# Conclusions

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- ▶ **First WW jet-inclusive differential measurements performed at LHC**
- ▶ **Reduction of uncertainties in dominant top background** using powerful data-driven  $t\bar{t}$  estimate
- ▶ **Fiducial and differential cross sections agree with theoretical predictions** up to highest measured  $p_T$  and for up to 5 jets

**Analysis team proceeding to look at WW + 0 jets**

In near future combine results for high precision fully inclusive measurement

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# Backup

- Number of  $t\bar{t}$  events passing  $e\mu$  selection:

$$N_{t\bar{t}} = \mathcal{L}\sigma_{t\bar{t}}\epsilon_{e\mu} \leftarrow e\mu \text{ selection efficiency}$$

$\uparrow$   
 $t\bar{t}$  (+jets) cross-section

- Number of  $t\bar{t}$  events in CRs obtained from data (backgrounds estimated with MC)

$$N_{t\bar{t}}^{2b} = \mathcal{L}\sigma_{t\bar{t}}\epsilon_{e\mu} \cdot C_b\epsilon_b^2$$

$$N_{t\bar{t}}^{1b} = \mathcal{L}\sigma_{t\bar{t}}\epsilon_{e\mu} \cdot 2\epsilon_b(1 - C_b\epsilon_b)$$

$\epsilon_b$  = efficiency to find and tag a  $b$ -jet  
 $\epsilon_{bb}$  = efficiency to find and tag two  $b$ -jets  
 $C_b = \epsilon_{bb}/\epsilon_b^2$  = correlation factor

- Solve for  $\epsilon_b = \frac{2N_{t\bar{t}}^{2b}}{C_b(N_{t\bar{t}}^{1b} + 2N_{t\bar{t}}^{2b})}$        $\mathcal{L}\sigma_{t\bar{t}}\epsilon_{e\mu} = \frac{C_b(N_{t\bar{t}}^{1b} + 2N_{t\bar{t}}^{2b})^2}{4N_{t\bar{t}}^{2b}}$

- Obtain estimate in SR

$$\begin{aligned}
 N_{t\bar{t}}^{SR} &= N_{t\bar{t}} - N_{t\bar{t}}^{1b} - N_{t\bar{t}}^{2b} \\
 &= \mathcal{L}\sigma_{t\bar{t}}\epsilon_{e\mu} \cdot (1 - 2\epsilon_b + C_b\epsilon_b^2)
 \end{aligned}$$

$$N_{t\bar{t}}^{SR} = \frac{C_b(N_{t\bar{t}}^{1b} + 2N_{t\bar{t}}^{2b})^2}{4N_{t\bar{t}}^{2b}} - N_{t\bar{t}}^{1b} - N_{t\bar{t}}^{2b}$$

$$N_{t\bar{t}}^{ib} = N_{\text{data}}^{ib} - N_{\text{bkg,MC}}^{ib}$$

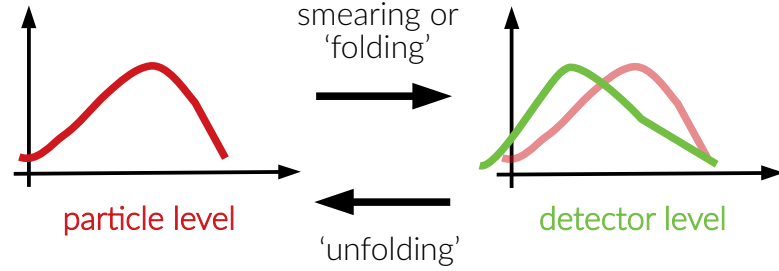
$$C_b = \frac{4N_{t\bar{t},\text{MC}}^{0+1+2b}N_{t\bar{t},\text{MC}}^{2b}}{\left(N_{t\bar{t},\text{MC}}^{1b} + 2N_{t\bar{t},\text{MC}}^{2b}\right)^2}$$

only use of  $t\bar{t}$  modelling  
 $C_b^{\text{fid.}} = 0.991 \pm 0.002$

# Unfolding details

- ▶ **Want fiducial-level cross sections** → need to 'undo' detector effects

→ D'Agostini, 95  
arXiv:1010.0632



- ▶ **Use iterative Bayesian unfolding method**

- Events can move from one truth-level bin to a different detector-level bin → describe by *migration matrix*
- Match truth and detector level selections as closely as possible
- Additional corrections applied for reconstruction inefficiencies/inaccuracies

Fiducial selection requirements	
$p_T^\ell$	> 27 GeV
$ \eta^\ell $	< 2.5
$m_{e\mu}$	> 85 GeV
$p_T^j$	> 30 GeV
$ y^j $	< 4.5

$$N_{\text{unf}}^j = \frac{1}{\underbrace{\epsilon_{\text{reco}}^j}_{\text{reco. efficiency}}} \sum_{i=1}^{n_{\text{reco bins}}} (N_{\text{data}}^i - N_{\text{bkg}}^i) \cdot \underbrace{f_{\text{fid}}^i}_{\text{fid. correction}} \cdot \underbrace{P(N_{\text{true}}^j | N_{\text{reco}}^i)}_{\text{calculated from migration matrix via Bayes' theorem}}$$

← contains prior from signal model

iterate

- ▶ **Optimise number of iterations** to reduce prior bias and stat. fluctuations

# Extra: EFT interpretation

- ▶ **SM can be considered as EFT** with additional dim.  $> 4$  operators suppressed by some UV scale  $\Lambda$

$$\mathcal{L}_{\text{SMEFT}} = \mathcal{L}_{\text{SM}} + \sum_{i,d>4} \frac{c_i^{(d)}}{\Lambda^{d-4}} Q_i^{(d)}$$

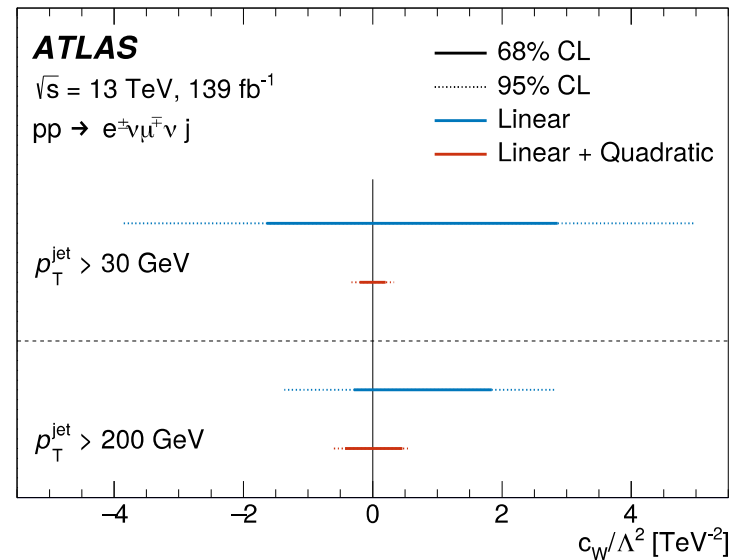
- ▶ **Small scale EFT study:** focus on dim. 6

- Analysis sensitive to  $Q_W$  affecting gauge boson self-couplings  $\longrightarrow$  [arXiv:1008.4884](https://arxiv.org/abs/1008.4884)
- Importance of SM+BSM interference term in cross section expected to increase with jet  $p_T$

$$\sigma = \underbrace{\sigma_{\text{SM}}}_{\text{purely SM}} + \underbrace{\frac{c_W}{\Lambda^2} \sigma_{\text{int}}}_{\text{SM+BSM}} + \underbrace{\frac{c_W^2}{\Lambda^4} \sigma_{\text{BSM}}}_{\text{purely BSM}}$$

## Fitting and results

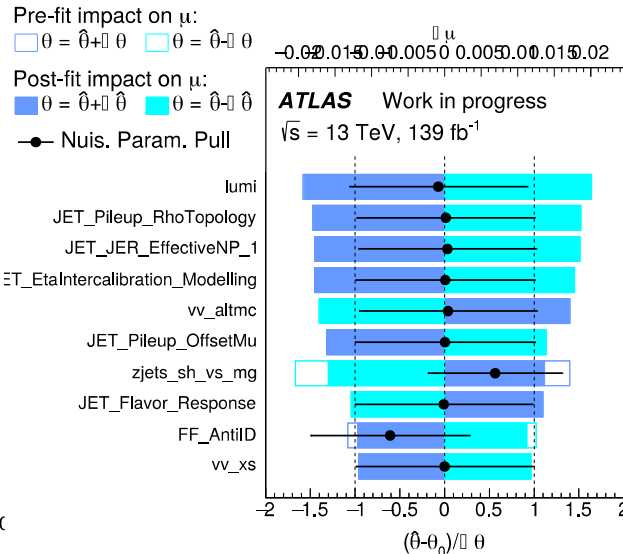
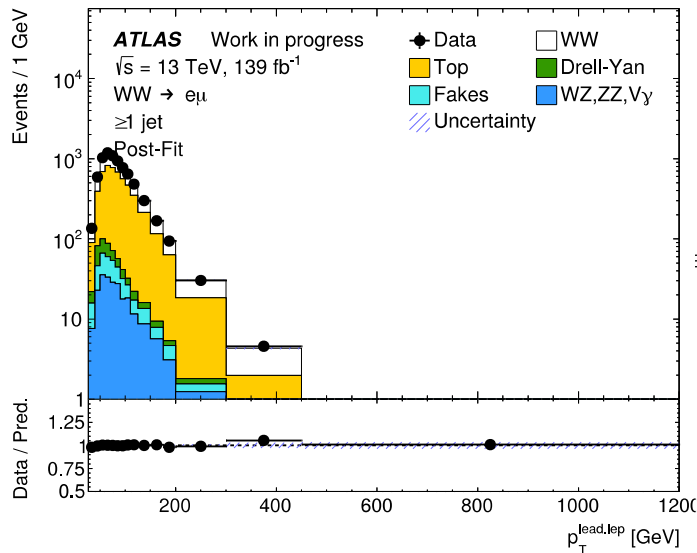
- ▶ **Perform likelihood fits in dedicated  $p_T^{\text{lead,jet}} > 200$  GeV region** using (unfolded)  $m_{e\mu}$  distribution
- ▶ **Impact of quadratic term seen to reduce** compared to  $p_T^{\text{lead,jet}} > 30$  GeV (nominal SR) fit



[arXiv:2103.10319](https://arxiv.org/abs/2103.10319)

# Looking to the future: fitting

- ▶ **Performing profile likelihood fits to data in SR** → alternative strategy to “cut-and-count” approach shown here
  - Use simulated signal and backgrounds on reconstruction-level, including effects of modelling and experimental systematic uncertainties
  - Extract signal (and background) normalisation as POI
- ▶ e.g. **validation of top estimate** (shown here) in  $p_T^{\text{lead,lep}}$  fit with  $\mu_{\text{WW}}$  and  $\mu_{\text{Top}}$  as POIs



$$\hat{\mu}_{\text{WW}} = 0.96 \pm 0.09$$

$$\hat{\mu}_{\text{Top}} = 1.00 \pm 0.05$$