

# ATLAS

## Boosted Object Tagging: Vector Boson

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## Motivation:

Including hadronically decaying boson increases sensitivity,  
especially because the hadronic decay has a large branching ratio

→ discriminate *boson-jets* from *QCD-jets*

Boosted region provides a better S/B ratio

W-tagging techniques can be extended to tag other bosons.

Additionally, the next natural step consists in distinguishing between W and Z  
(WZ/WW separation, BSM diboson, leptophobic FCNC)

## Content:

- **W-tagging in Run1**  
(Grooming and Substructure variables)
- **W- and Z-tagging in Run 2**
- **W/Z separation in Run 1**

PERF-2015-03

ATL-PHYS-PUB-2015-033

PERF-2015-02



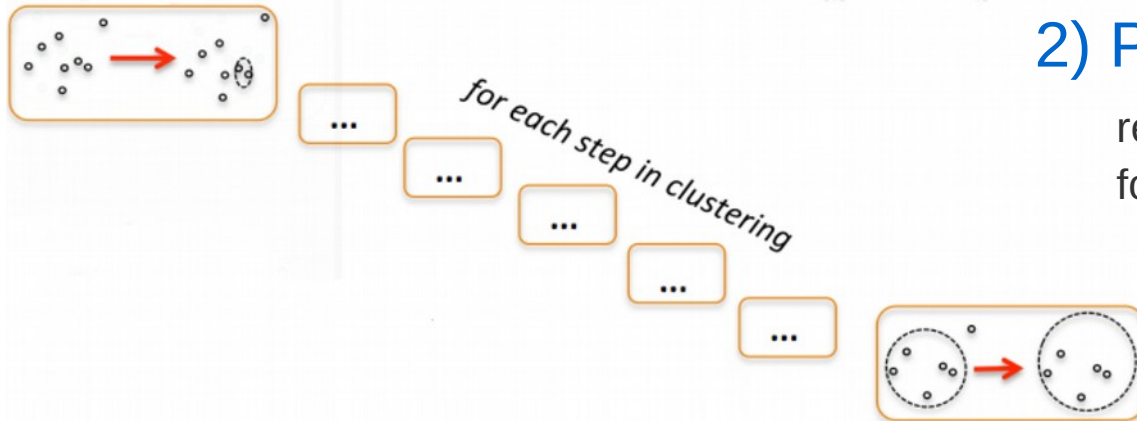
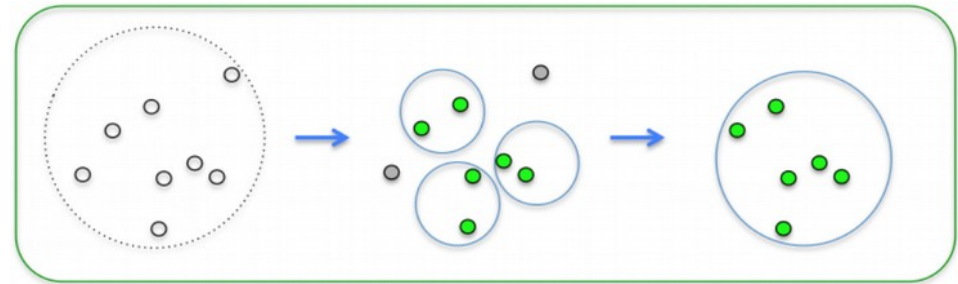
# W-tagging in Run1

Material from **PERF-2015-03**



## 1) Trimming

reclustering: kt algorithm, with  $R=R_{sub}$   
 If  $p_T(\text{subject}) < f_{cut} \times p_T(\text{jet}) \rightarrow$  discard subject



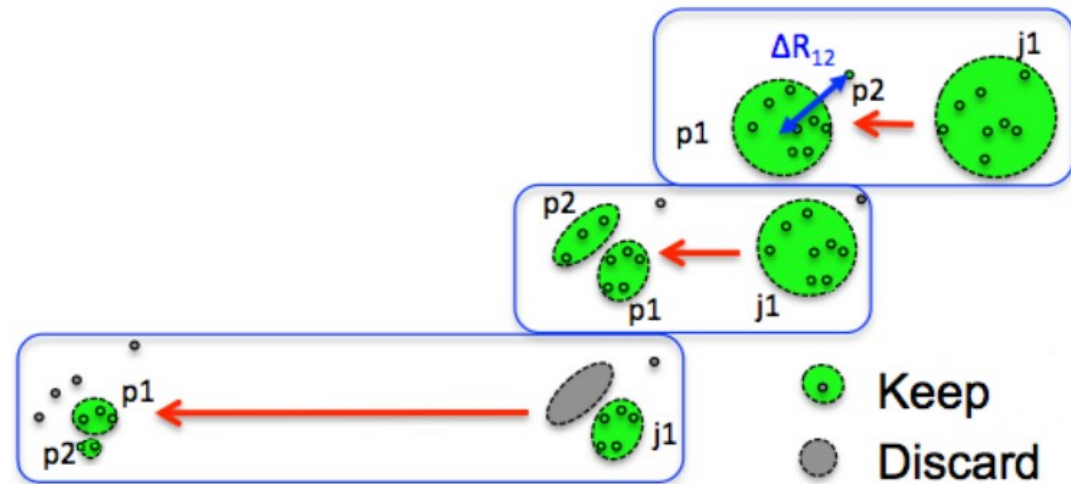
## 2) Pruning

reclustering: C/A algorithm  
 for pair  $ij$  with  $p_T(i) < p_T(j)$ , if  $\Delta R_{ij} > R_{cut} \times 2m/p_T$   
 or  $p_T(i)/p_T(j) < z_{cut}$   
 $\rightarrow$  discard object  $i$

## 3) Split-Filtering (mass-drop, BDRS)

1) Splitting:  
 de-clustering C/A algorithm  
 stop when  $\max(m_1, m_2)/m_{12} < \mu_{frac}$  and  $y_{12} > y_{cut}$   
 otherwise, discard lowest mass object

2) Filtering:  
 reclustering with  $R=R_{sub}$   
 only the 3 hardest subjects are kept





Three grooming techniques have been studied → >500 jet collections

## 1) Trimming

### Trimming configurations

| Input jet algorithms | $R$                | $R_{\text{sub}}$ | $f_{\text{cut}}$ (%)            |
|----------------------|--------------------|------------------|---------------------------------|
| C/A, anti- $k_t$     | 0.6, 0.8, 1.0, 1.2 | 0.1, 0.2, 0.3    | 1, 2, 3, 4, 5, 7, 9, 11, 13, 15 |

$$\rightarrow 2 \times 4 \times 3 \times 10 = 240$$

## 2) Pruning

### Pruning configurations

| Input jet algorithm | $R$           | Reclust. alg. | $Z_{\text{cut}}$ (%) | $R_{\text{cut}}$  |
|---------------------|---------------|---------------|----------------------|---|
| C/A, anti- $k_t$    | 0.8, 1.0, 1.2 | C/A           | 10, 15, 20, 25, 30   | $\frac{1}{100}, \frac{1}{10}, \frac{1}{8}, \frac{1}{4}, \frac{1}{2}, 1.0$ |

$$\rightarrow 2 \times 3 \times 1 \times 5 \times 6 = 180$$

## 3) Split-Filtering

### Split-filtering configurations

| Input jet algorithm | $R$           | $R_{\text{sub}}$             | $\mu_{\text{frac}}$ | $y_{\text{cut}}$      |
|---------------------|---------------|------------------------------|---------------------|-----------------------|
| C/A                 | 0.8, 1.0, 1.2 | 0.3, $\min(0.3, \Delta R/2)$ | 67, 78, 89, 100     | 0.06, 0.07, ..., 0.20 |

$$\rightarrow 1 \times 3 \times 2 \times 4 \times 11 = 264$$



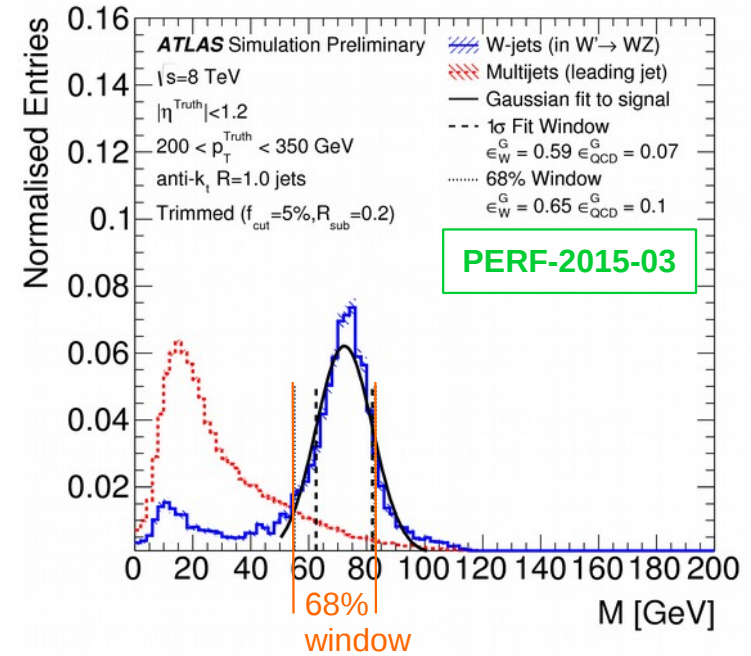
# Grooming 3/4

## Study of the grooming techniques

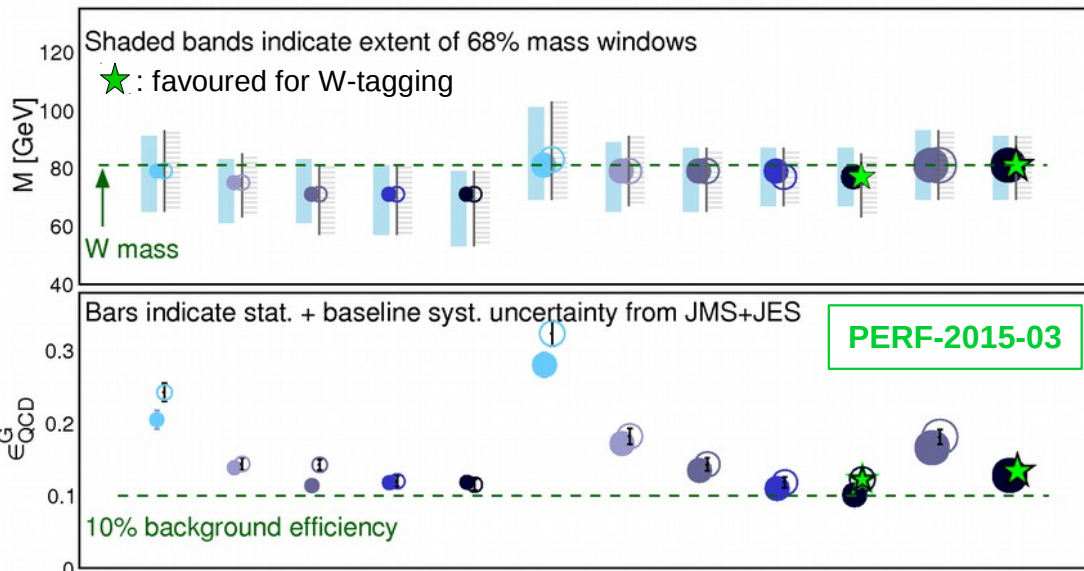
Smallest mass windows containing 68% of signal are defined  
 For 3 region of truth-jet  $p_T$  (200 - 350 - 500 - 1000 GeV)

Reject pathological cases:

- the 68% mass window contains  $m_W$
- mass peak is too asymmetric
- background in the window has irregular shape
- not sufficiently stable w.r.t. Pile-Up



Trimmed jets,  $R=1.0$  C/A  $R_{sub} = 0.2$   $f_{cut} = 3\%$   $\sqrt{s} = 8$  TeV  
 $350 < p_T^{Truth} < 500$  GeV anti- $k_t$   $R_{sub} = 0.2$   $f_{cut} = 3\%$   $\sqrt{s} = 8$  TeV  
 $|\eta^{Truth}| < 1.2$   $R_{sub} = 0.2$   $f_{cut} = 3\%$   $\sqrt{s} = 8$  TeV



Sort by best QCD multijets  
 background rejection  
 in the mass window

← example (trimming)

→ 27 best jet collections are kept

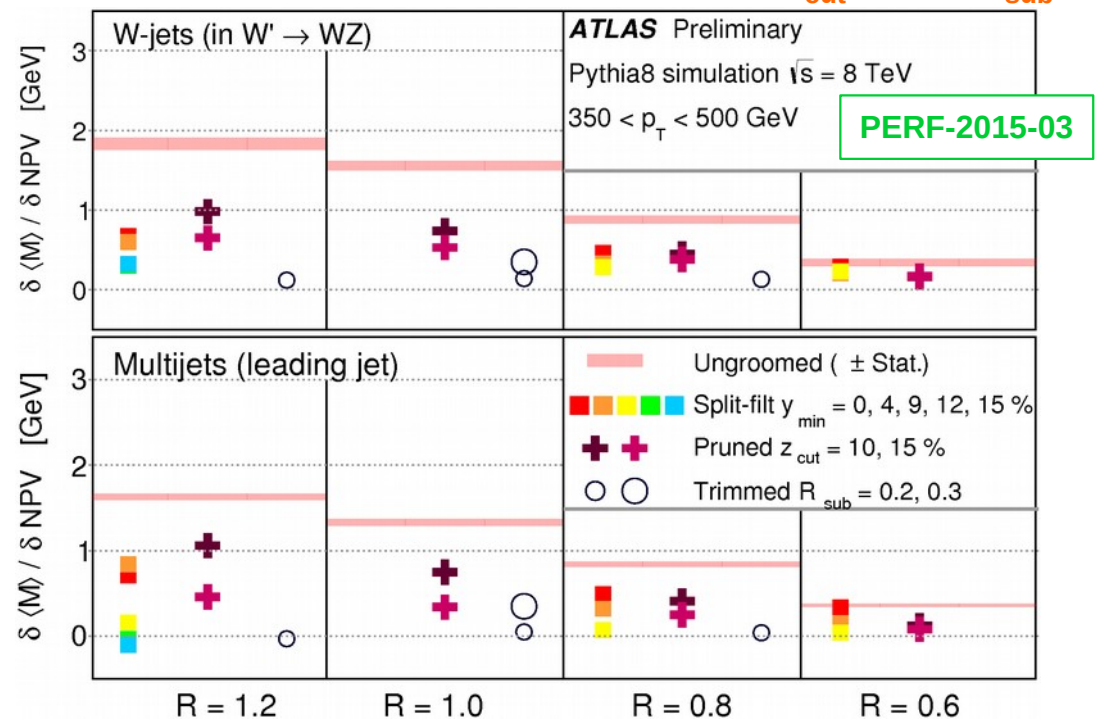
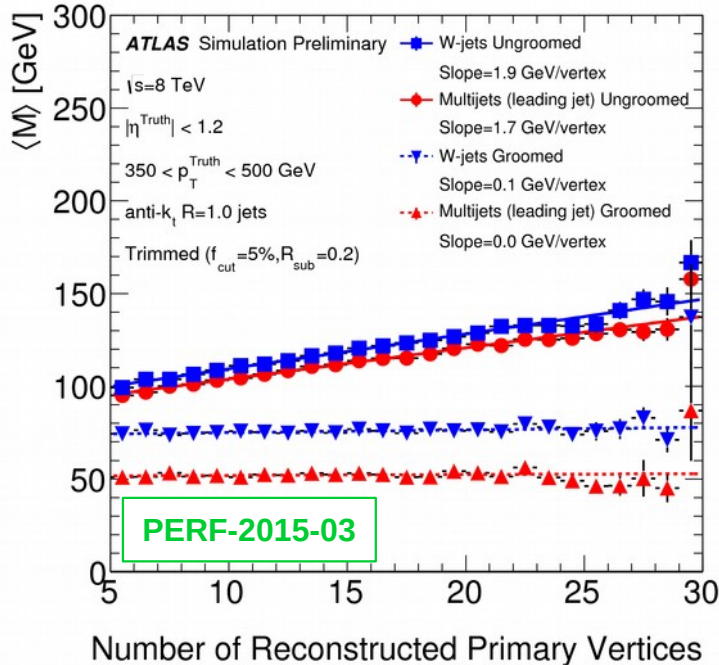


# Grooming 4/4

## Study observations:

- $R=0.6$  or  $R=0.8$  are too small (loss of constituents for  $p_T < 500$  or  $350\text{GeV}$  respectively)
- For high  $p_T$ , performances of the different grooming techniques are similar
- For **trimming**, C/A and anti-kt are similar, and best for large values of  $f_{\text{cut}}$
- For **pruning**, C/A always better than kt, and for low  $p_T$ , other grooming methods preferred
- For **split-filtering**, larger  $y_{\text{cut}}$  is better, optimal param. depends of  $p_T$ , but less dependence with  $p_T$  and other param for  $y_{\text{cut}} > 0.09$
- Grooming reduces Pile-Up dependency (so, no additional P-U removal procedure needed)

→ best grooming: low bkgd eff. + good P-U stability → **anti-kt  $R=1.0$  trimmed  $f_{\text{cut}}=5\%$ ,  $R_{\text{sub}}=0.2$**





# Substructure variables 1/3



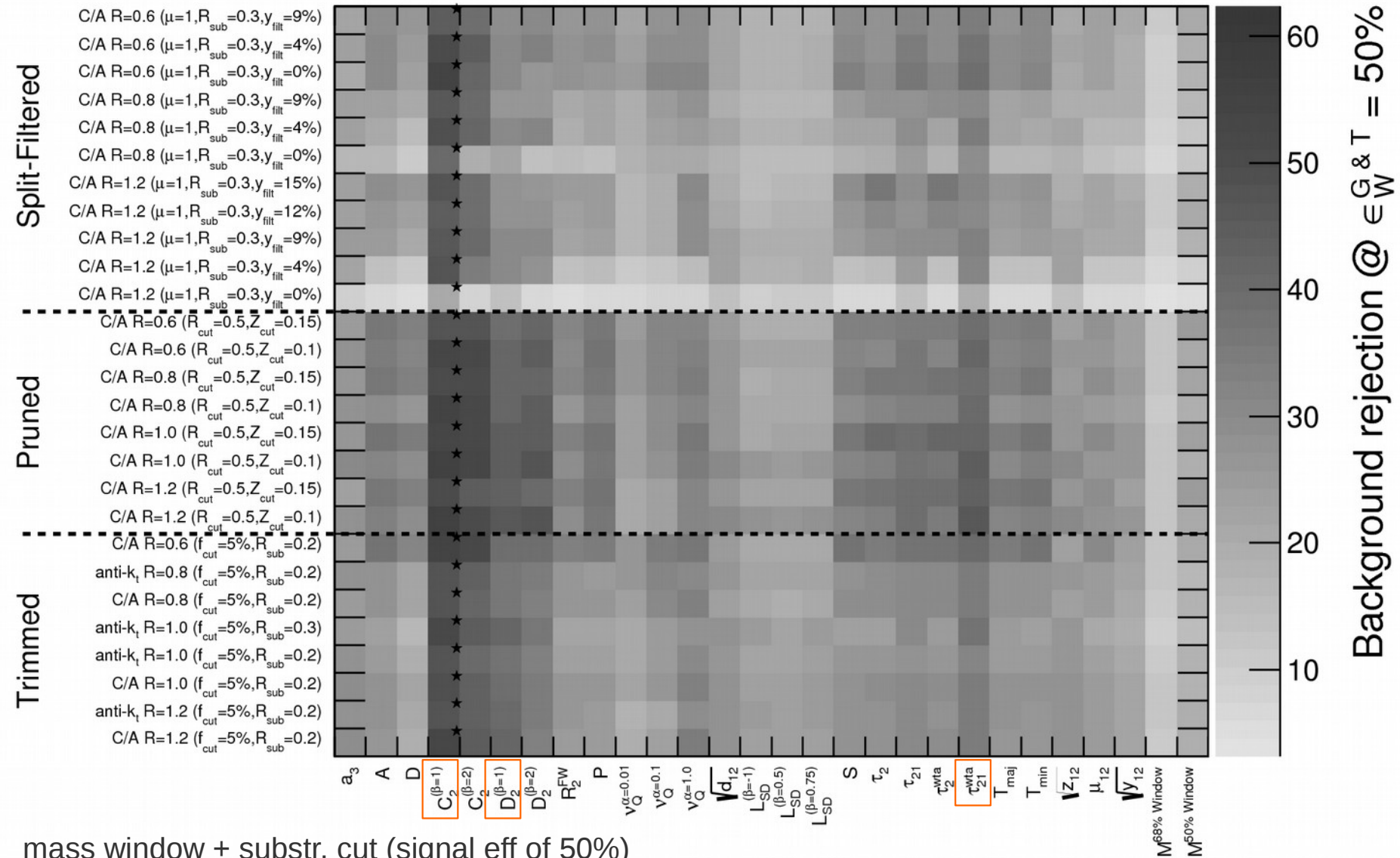
**ATLAS** Simulation Preliminary Jet 4-momentum not calibrated

$\sqrt{s}=8$  TeV

$|\eta^{\text{Truth}}| < 1.2$ ,  $350 < p_T^{\text{Truth}} < 500$  GeV, M Cut

26 substructure variables studied

★ = Optimal substructure variable for jet algorithm



PERF-2015-03

mass window + substr. cut (signal eff of 50%)





# Substructure variables 2/3

## Best variables:

For the constituents  $i$  inside the large-R jet  $J$

Mass:  $M^2 = (\sum_i E_i)^2 - (\sum_i p_i)^2$

$\tau_{21}^{wta}$ :  $\tau_0(\beta) = \sum_{i \in J} p_{T_i} \Delta R^\beta$ ,

$\tau_N$ : exactly N subsets kt reconstructed

$a_i$ : axis of the  $i^{th}$  subset (if Winner-Take-All (wta), axis of the hardest constituent)

$\tau_1(\beta) = \frac{1}{\tau_0(\beta)} \sum_{i \in J} p_{T_i} \Delta R_{a_1, i}^\beta$ ,

here:  $\beta = 1$

$\tau_2(\beta) = \frac{1}{\tau_0(\beta)} \sum_{i \in J} p_{T_i} \min(\Delta R_{a_1, i}^\beta, \Delta R_{a_2, i}^\beta)$ ,

$\tau_{21} = \frac{\tau_2}{\tau_1}, \tau_{21}^{wta} = \frac{\tau_2^{wta}}{\tau_1^{wta}}$

## $C_2^{(\beta=1)}$ and $D_2^{(\beta=1)}$ :

$E_{CF0}(\beta) = 1$ ,

$E_{CF1}(\beta) = \sum_{i \in J} p_{T_i}$ ,

$E_{CF2}(\beta) = \sum_{i < j \in J} p_{T_i} p_{T_j} (\Delta R_{ij})^\beta$ ,

$E_{CF3}(\beta) = \sum_{i < j < k \in J} p_{T_i} p_{T_j} p_{T_k} (\Delta R_{ij} \Delta R_{ik} \Delta R_{jk})^\beta$ ,

$e_2^{(\beta)} = \frac{E_{CF2}}{E_{CF1}^2}$ ,

$e_3^{(\beta)} = \frac{E_{CF3}}{E_{CF1}^3}$

$C_2^{(\beta)} = \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^2}$ ,

$D_2^{(\beta)} = \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^3}$

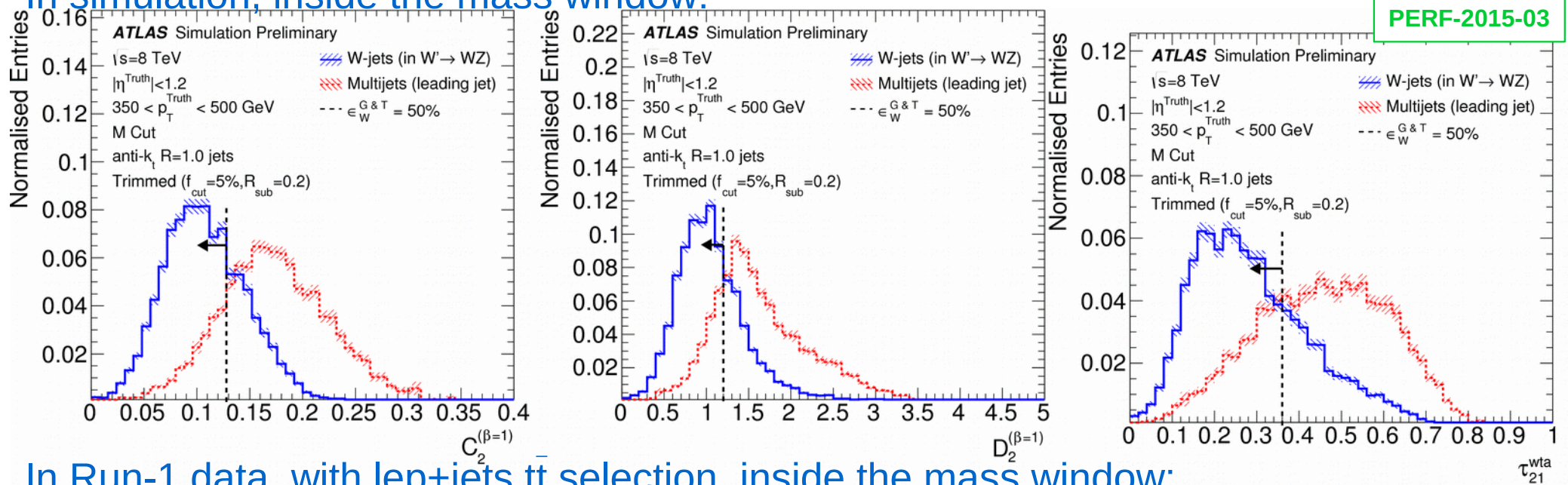
→ W-Tagger defined as grooming + mass window + cut on one substructure variable



# Substructure variables 3/3

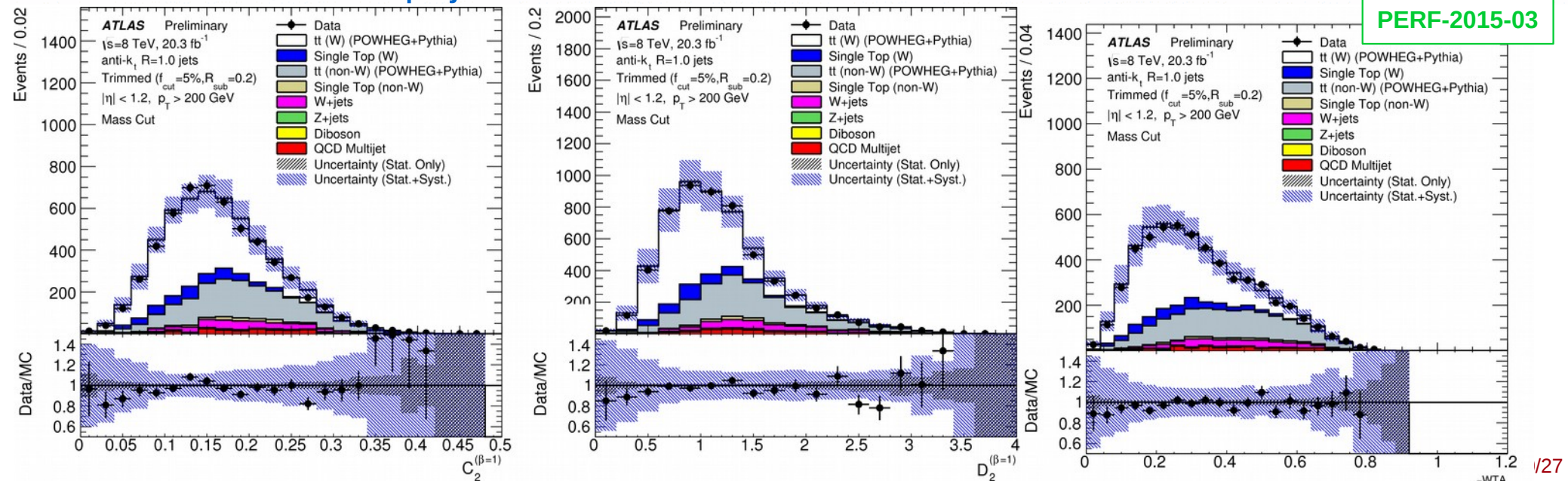
In simulation, inside the mass window:

PERF-2015-03



In Run-1 data, with lep+jets tt selection, inside the mass window:

PERF-2015-03





# W-tagging efficiencies 1/2

Two W-tagger working points: medium (50% signal eff.), tight (25% signal eff.)

Efficiency definition:

$$\epsilon_{MC} = \frac{N_{\text{Boosted-W}}^{\text{tagged}}}{N_{\text{Boosted-W}}^{\text{pre-tagged}}} \quad \epsilon_{\text{Data}} = \frac{N_{\text{tagged}} - N_{\text{Bkg}}^{\text{tagged}}}{N_{\text{pre-tagged}} - N_{\text{Bkg}}^{\text{pre-tagged}}}$$

Boosted-W = jet matched ( $\Delta R < 1.0$ ) with both W decay partons

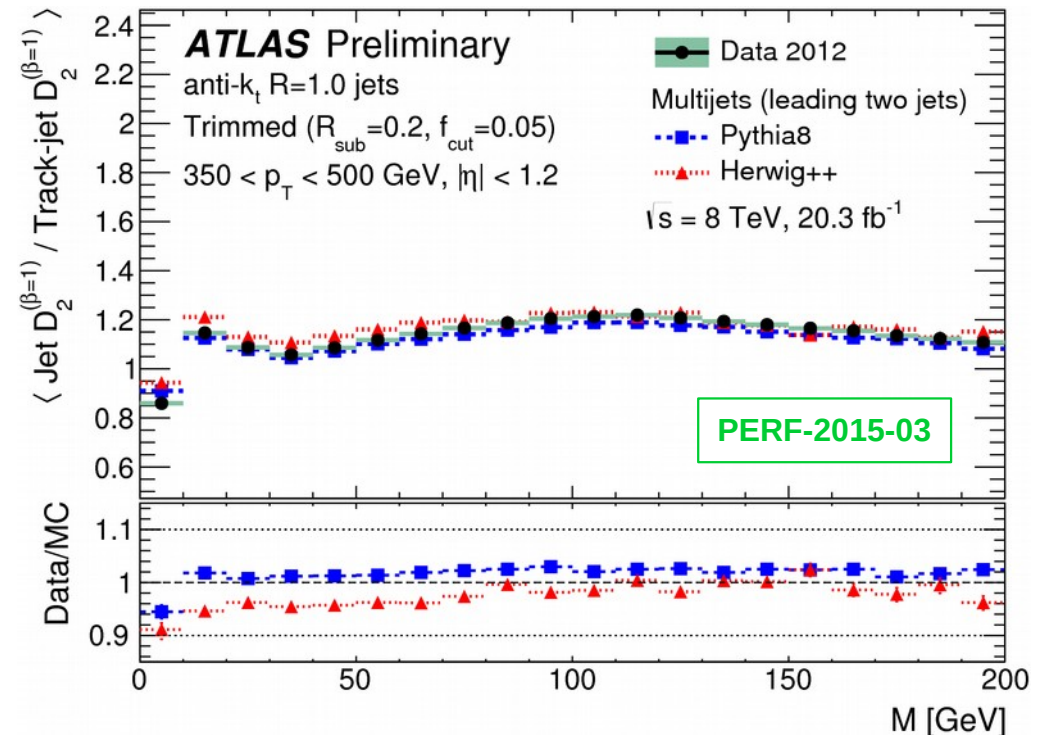
Bkg = multijet QCD and W+Jets from data-driven methods, others from simulation

## Systematic uncertainties:

- Large-R jets mass and energy scale/resolution
- Substructure variables (~5%)
- MC generator, ISR/FSR
- Backgrounds normalisation  
→ 6-13%

Example for  $D_2^{(\beta=1)}$  tagging with akt 1.0 trimming  $f_{\text{cut}}=5\%$ ,  $R_{\text{sub}}=0.2$

$D_2^{(\beta=1)}$  unc. from data-driven track-jet double-ratio method

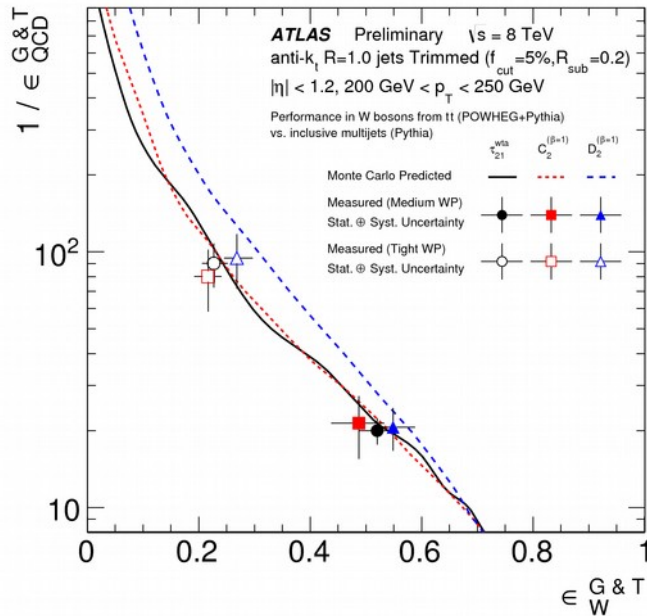




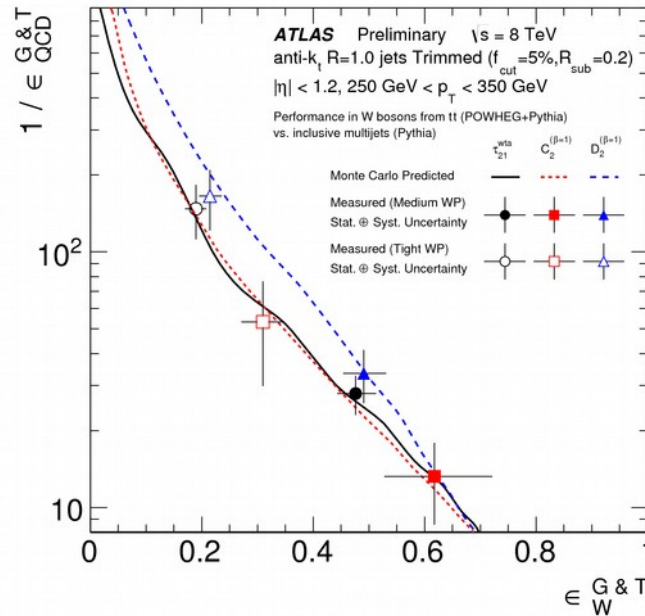
# W-tagging efficiencies 2/2

## Two W-tagger working points:

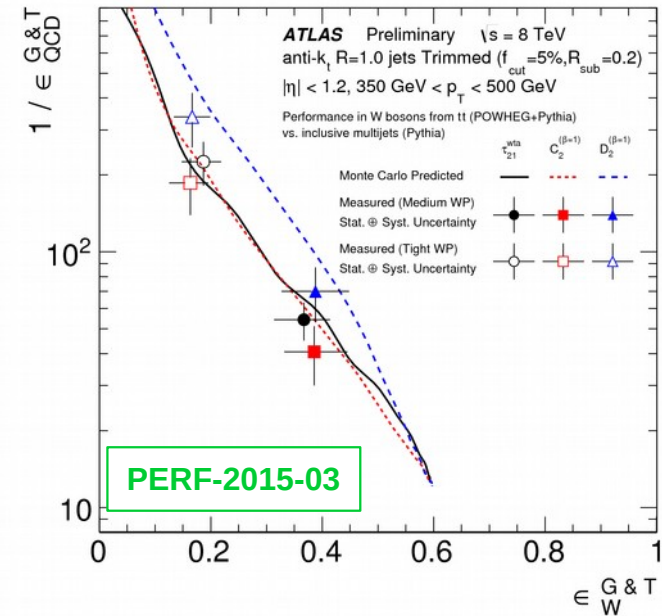
for anti-kt R=1.0 trimmed  $f_{\text{cut}}=5\%$ ,  $R_{\text{sub}}=0.2$ , cut on substr. variable



200 <  $p_T$  < 250 GeV



250 <  $p_T$  < 350 GeV



350 <  $p_T$  < 500 GeV

Data points are from fits using templates from Powheg+Pythia  
Compared with simulation scan

- Rem.:
- data from  $t\bar{t}$  topology, while the working-points have been optimized in W' topology
  - different MC generators (Powheg+Pythia vs. MC@NLO + Herwig) lead to different shapes



# W- and Z-tagging in Run2

Material from

**ATL-PHYS-PUB-2015-033**



# Strategy

## Early W-tagger for early Run2 analyses:

similar approach than in Run1:  
grooming + substructure cut

based on lesson from Run1:  
selected number of grooming techniques (4 instead of 27)

selected number of substructure variables (3:  $\tau_{21}^{wta}$ ,  $C_2^{(\beta=1)}$ ,  $D_2^{(\beta=1)}$ )

ATL-PHYS-PUB-2015-033

ATLAS Simulation Preliminary

$\sqrt{s} = 13$  TeV  $\star =$  Optimal grooming + tagging combination

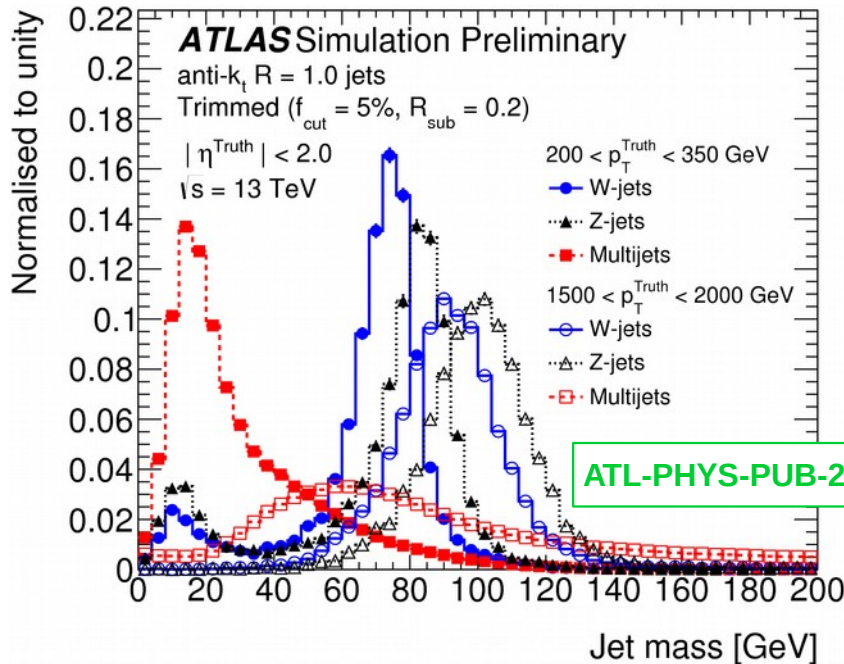
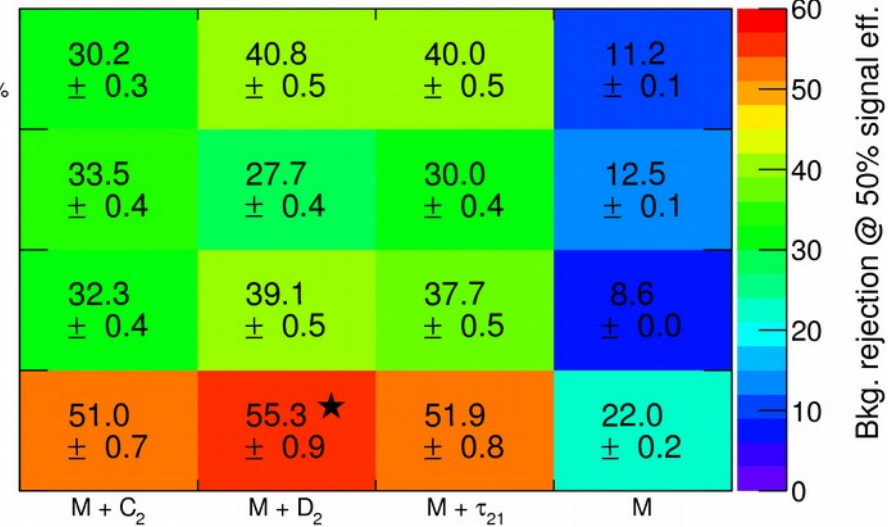
$|\eta^{\text{Truth}}| < 2.0, 200 < p_T^{\text{Truth}} < 350$  GeV,  $M^{\text{Reco}}$  Cut W-jets

C/A R = 1.0 jets  
Pruned  
 $R_{\text{cut}} = 0.5, z_{\text{cut}} = 15\%$

C/A R = 1.2 jets  
Split-Filtered  
 $\mu = 1.0, y_{\text{filt}} = 15\%$

C/A R = 1.2 jets  
Split-Filtered  
 $\mu = 1.0, y_{\text{filt}} = 4\%$

anti- $k_t$  R = 1.0 jets  
Trimmed  
 $f_{\text{cut}} = 5\%, R_{\text{sub}} = 0.2$



For W and Z-tagger

Syst. unc. from Run1 + Run2 MC

Final tagger based on:

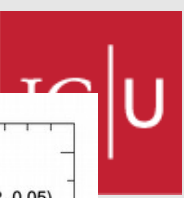
anti-kt R=1.0 trimmed  $f_{\text{cut}} = 5\%, R_{\text{sub}} = 0.2$

$D_2^{(\beta=1)}$  variable

$p_T$  parametrisation of the cut on  $D_2^{(\beta=1)}$



# Variables in Run2

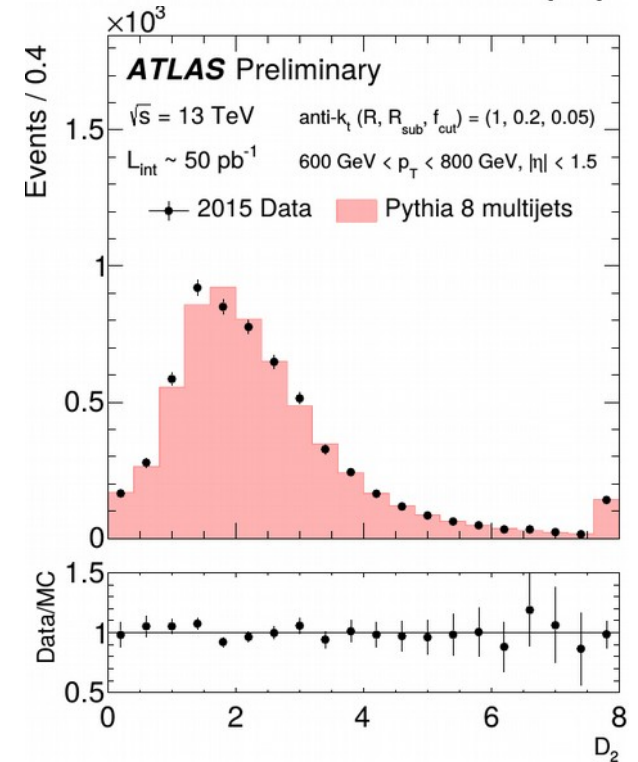
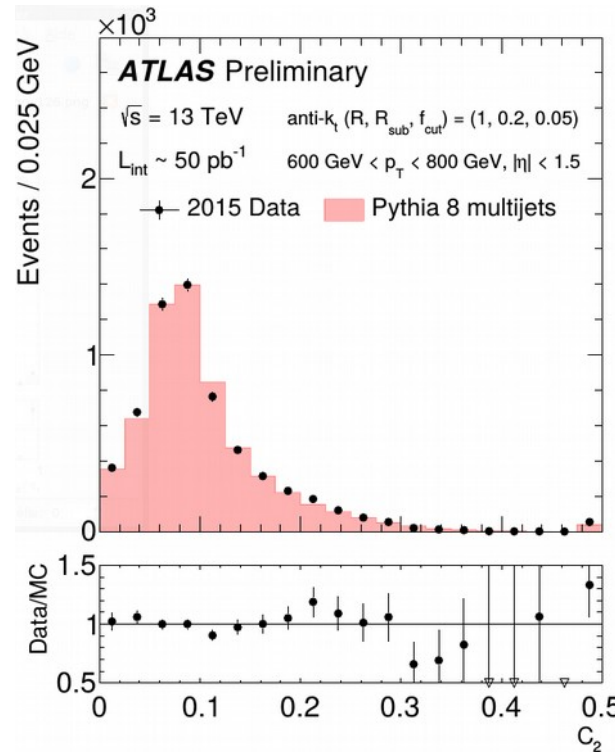
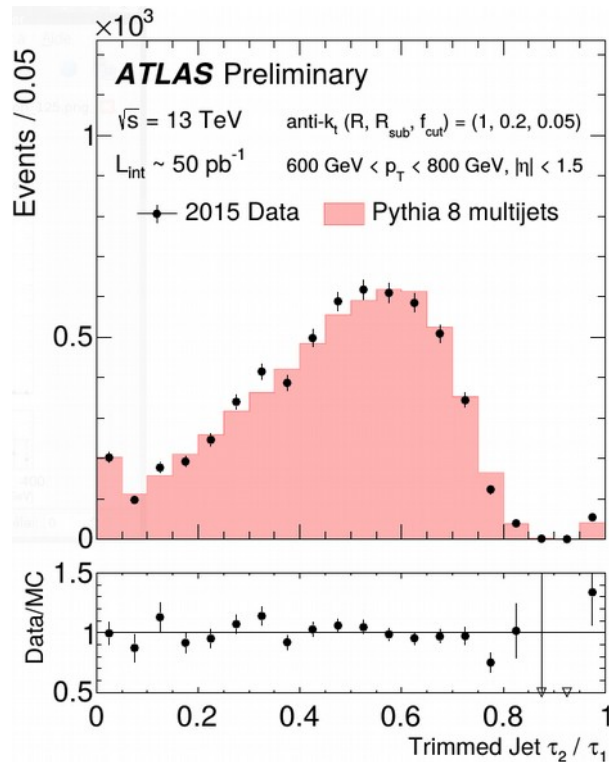
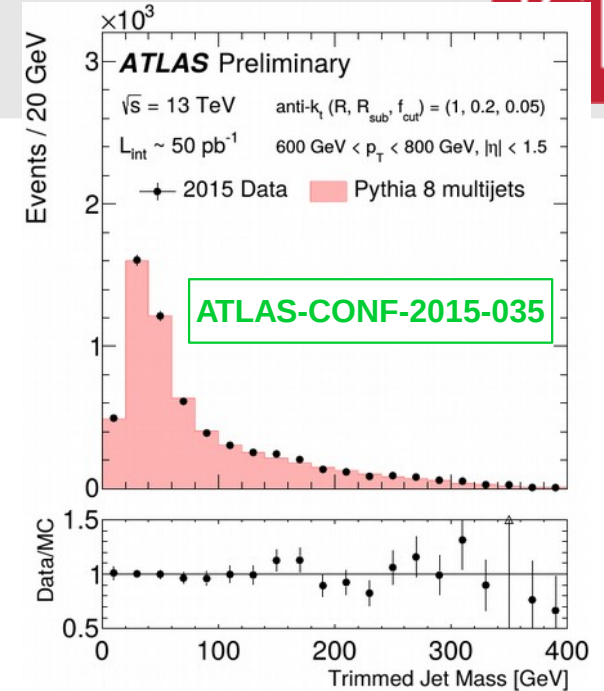


## First data in Run2:

$50 \text{ pb}^{-1}$ ,  $\sqrt{s} = 13 \text{ TeV}$

Anti-kt  $R=1.0$  trimmed  $f_{\text{cut}}=5\%$   $R_{\text{sub}}=0.2$

mass,  $\tau_{21}^{\text{wta}}$ ,  $C_2^{(\beta=1)}$ ,  $D_2^{(\beta=1)}$





# Systematic uncertainties

Uncertainties on  $p_T$ ,  $m$  and  $D_2^{(\beta=1)}$  :  $p_T < 1\text{TeV}$ : 3 Run2 variations + Run1 uncert. (quad. sum)  
 $p_T > 1\text{TeV}$ : 7 Run2 variations + Stat. uncert. (quad. sum)

Run2 variations: (uncert. = deviation from nominal in truth-reco jet response)

**Noise thr.** : noise threshold variation in the calorimeter cells clustering procedure

**QGSP\_BIC (FTFP\_BIC)**: Quark Gluon String model (Fritiof String model) + Binary Intranuclear Cascade

**Rescat.**: High-energy re-scattering simulated with Binary Intranuclear Cascade

**50ns vs 25ns:**

bunch spacing  
condition var.

**Herwig++:**

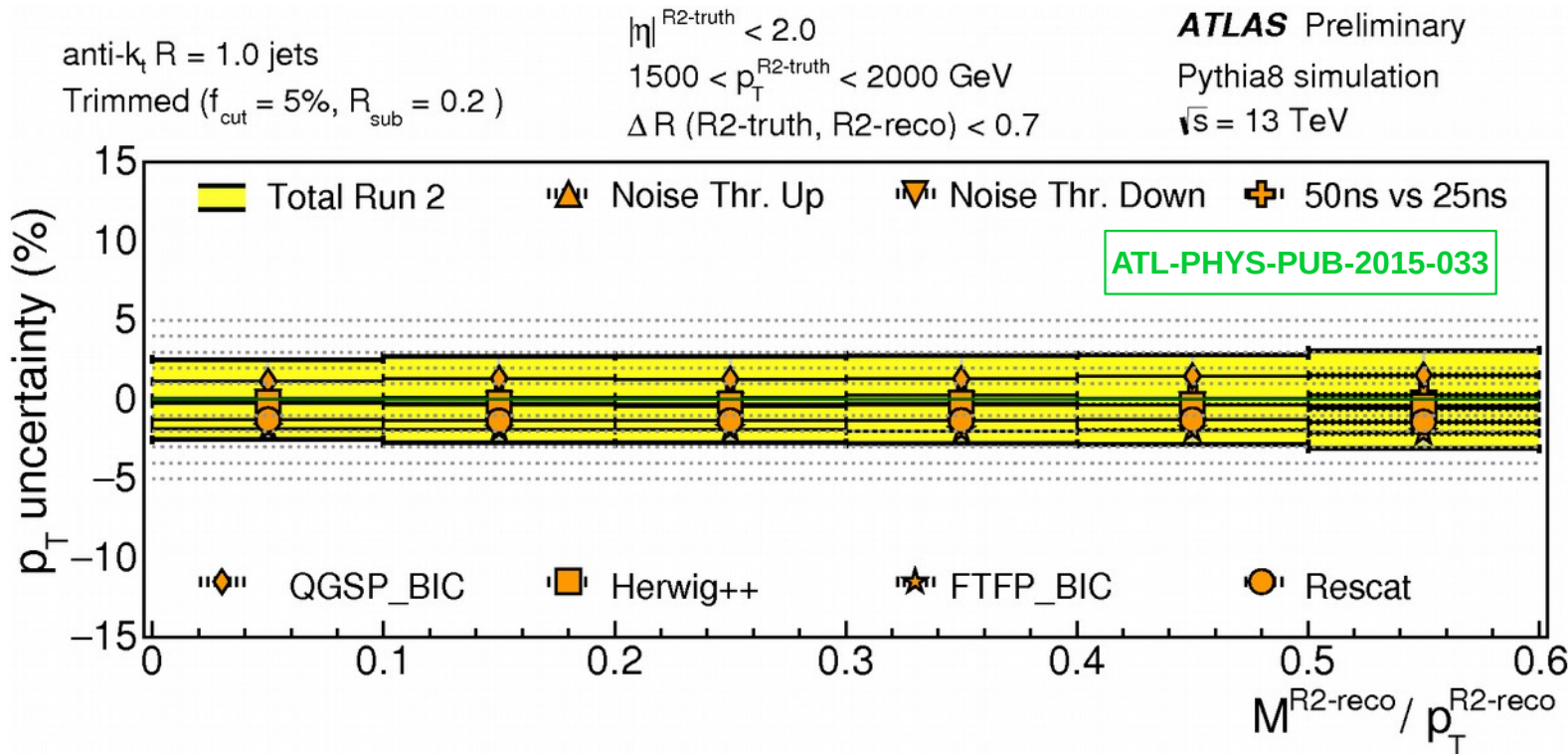
Herwig++ generator

For  $p_T < 1\text{TeV}$ :

Run1 from

track-jet double-ratio

data-driven method







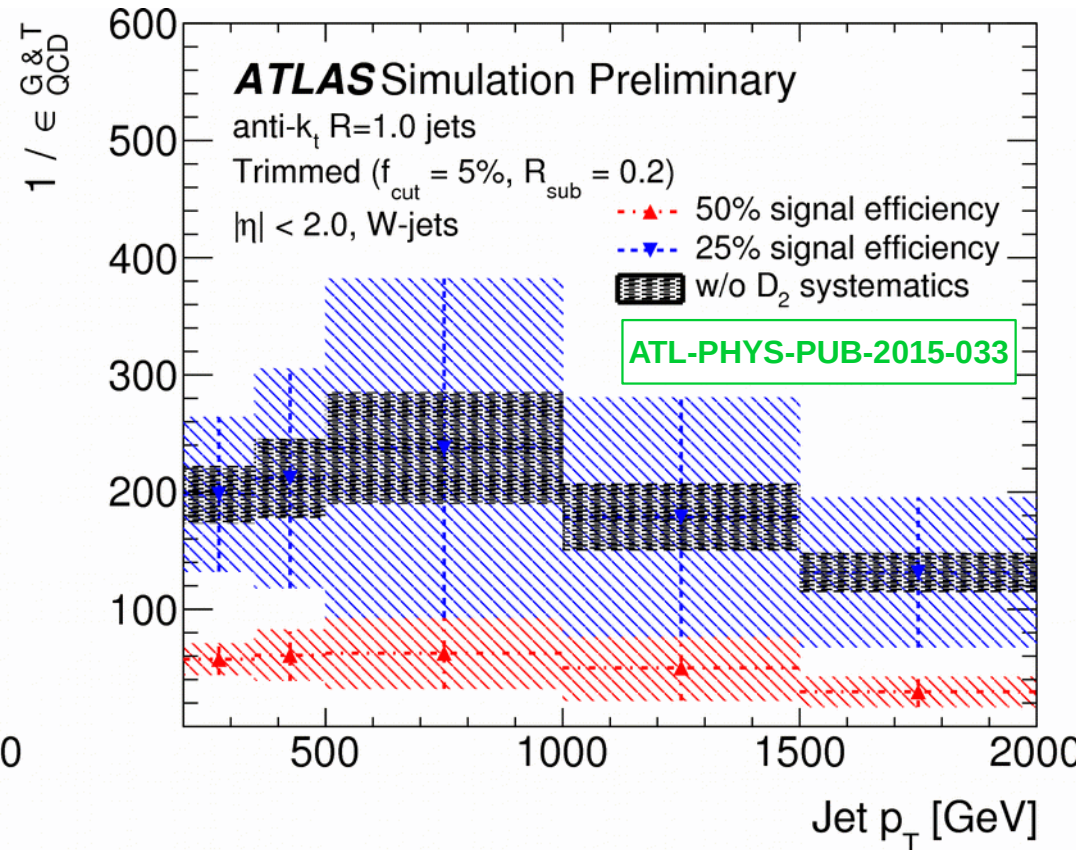
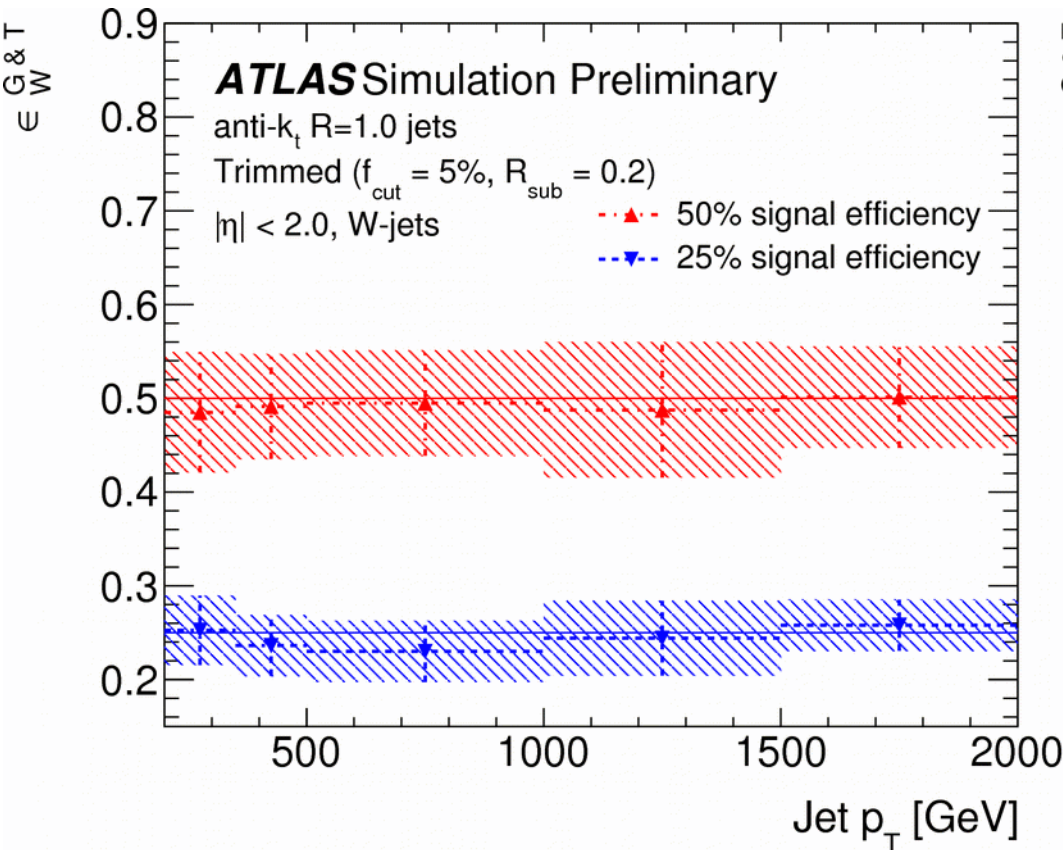
# Efficiencies 1/2

Two W-tagger working points: medium (50% signal eff.), tight (25% signal eff.)

Similar results available for Z-tagging (next slide)

Signal efficiencies and bkgd rejection, with respect to reco jet  $p_T$

Larger uncertainties at larger  $p_T$ , usually from  $D_2$  uncertainties





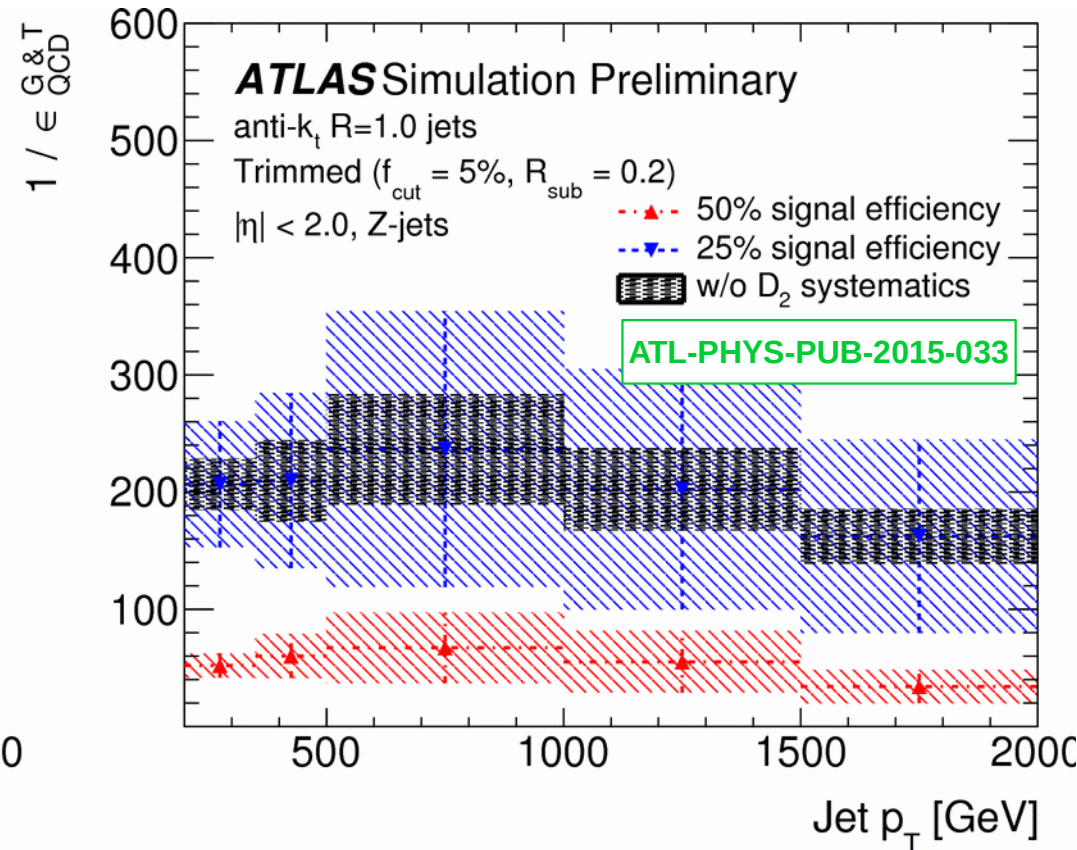
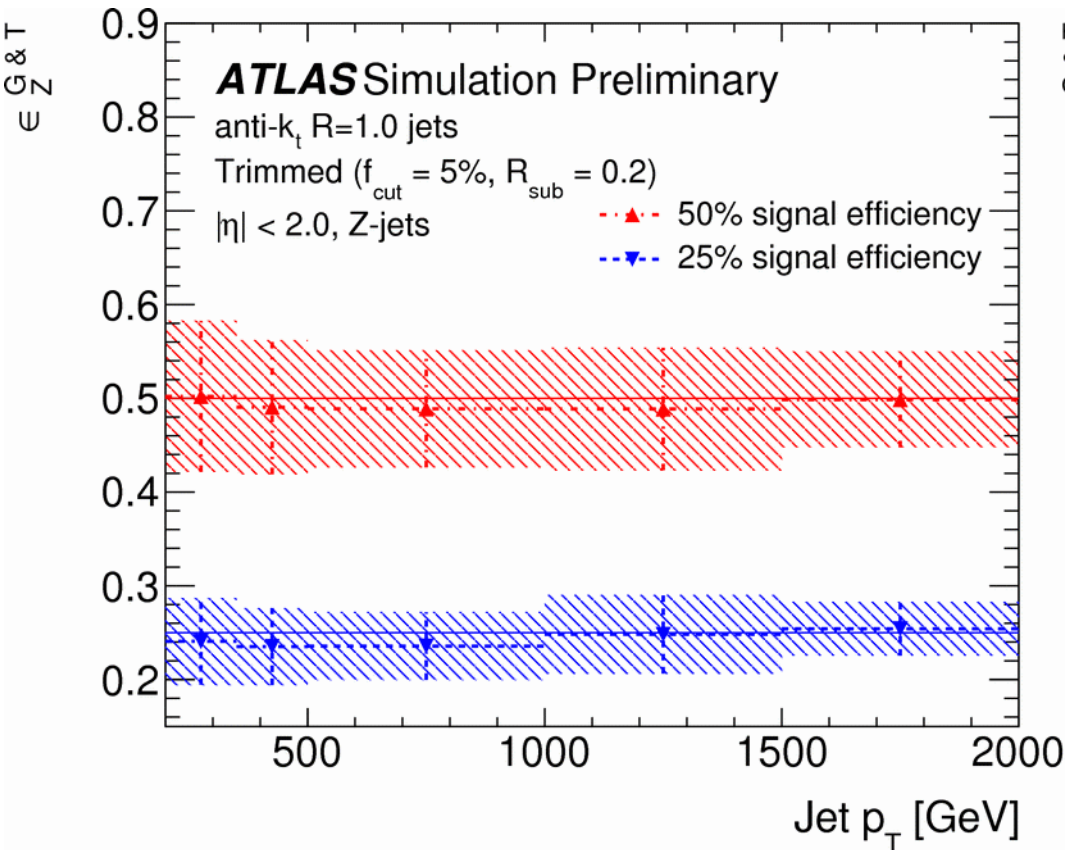
# Efficiencies 2/2

Two Z-tagger working points: medium (50% signal eff.), tight (25% signal eff.)

Similar results available for W-tagging

Signal efficiencies and bkgd rejection, with respect to reco jet  $p_T$

Larger uncertainties at larger  $p_T$ , usually from  $D_2$  uncertainties





# W/Z separation in Run1

Material from **PERF-2015-02**

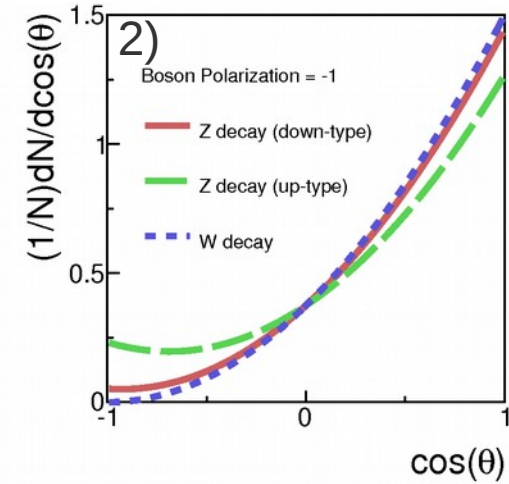
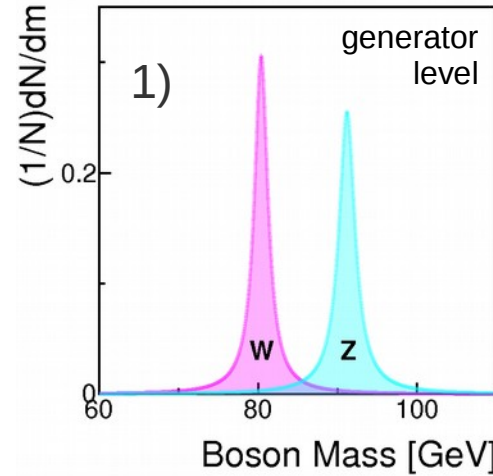


# W/Z discriminating variables 1/2

Use of substructure variables to distinguish hadronically decaying W and Z

## Differences between W and Z:

- 1) Mass
- 2) ~~Angular distributions~~ ← small at parton level, likely to be washed out by combinatorics, non-perturbative physics and reconstruction
- 3) Flavor decay ratio
- 4) Boson charge

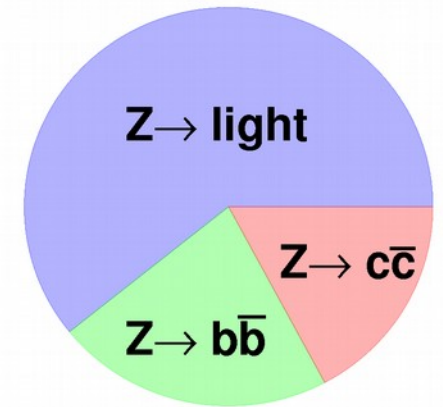
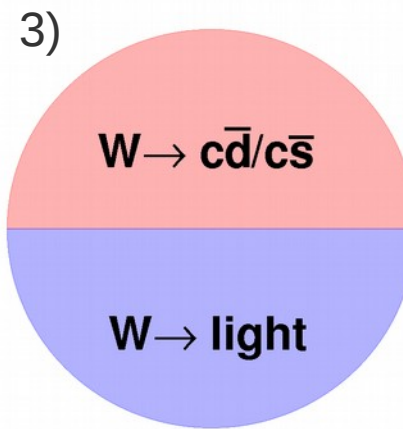


The boson is contained in a large-R jet:

anti-kt R=1.0, kt trimmed  $f_{cut}=5\%$ ,  $R_{sub}=0.3$  calibrated

This large-R jet can be matched with small-R jets:

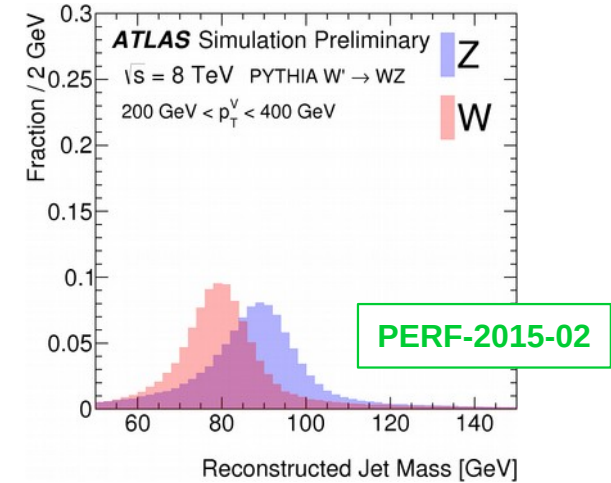
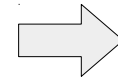
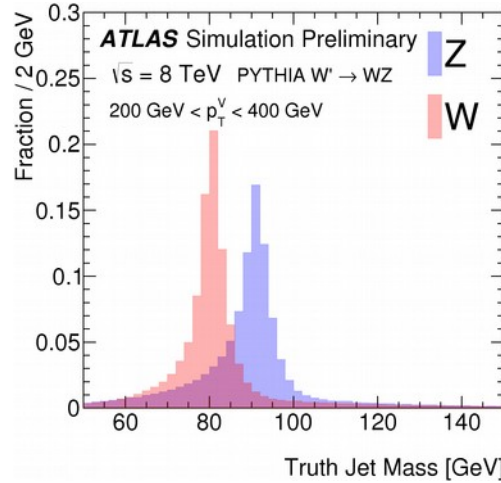
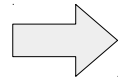
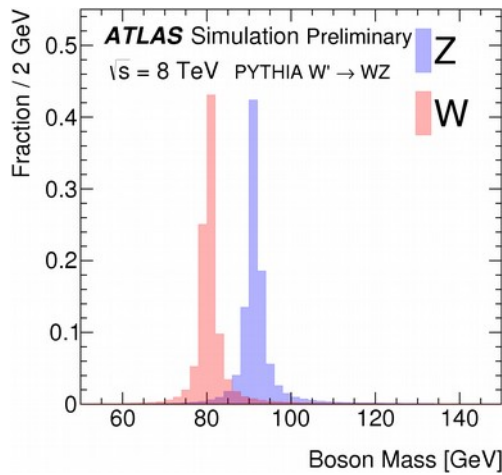
anti-kt R=0.4, matched with  $\Delta R(j,J)<1.0$  calibrated





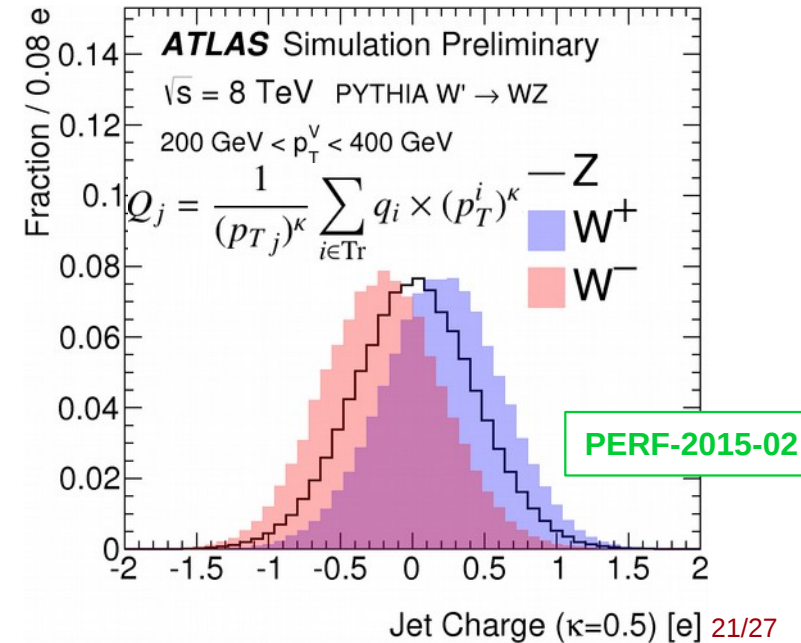
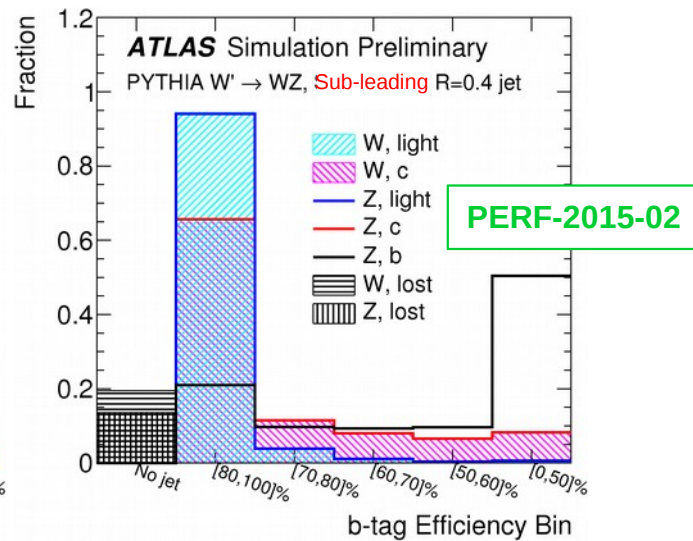
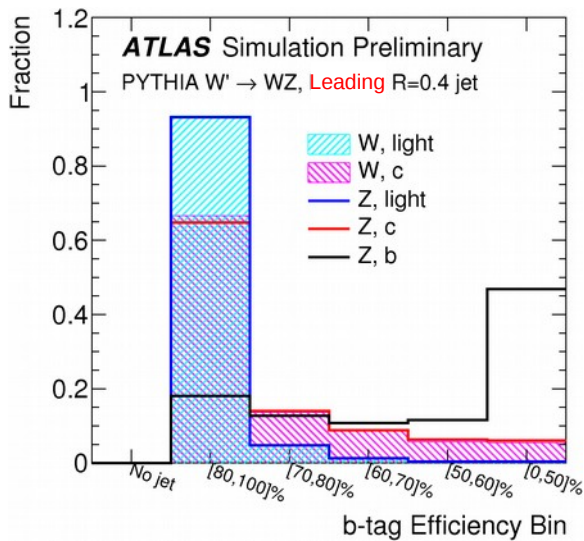
# W/Z discriminating variables 2/2

Mass resolution is degraded by parton fragmentation and reconstruction



2 first leading small jets matched with large jet, b-tagging (MV1)

Charge (cf. B. Nachman's talk)





# Comparison with Data

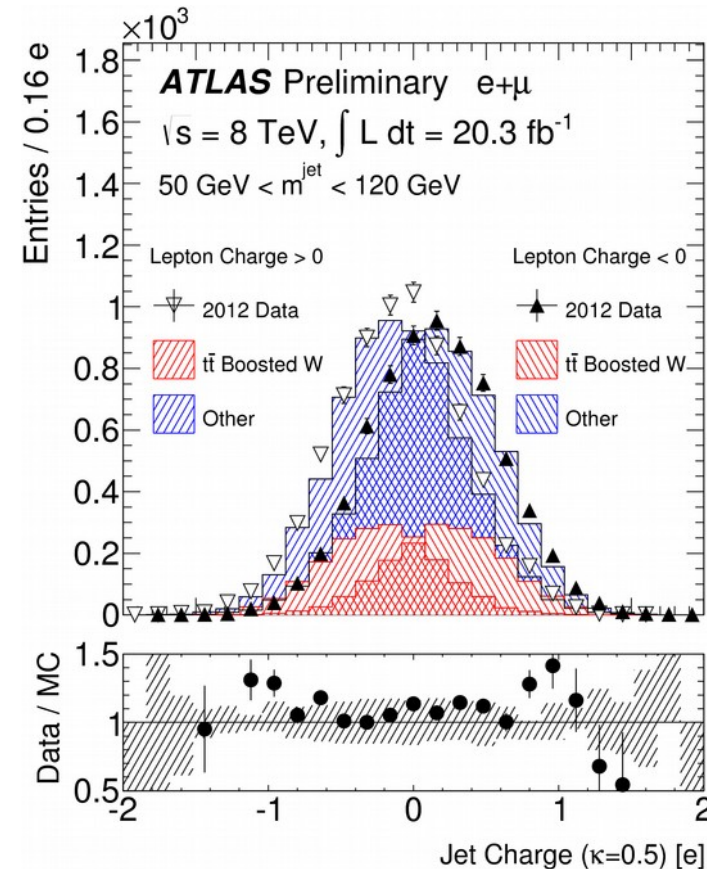
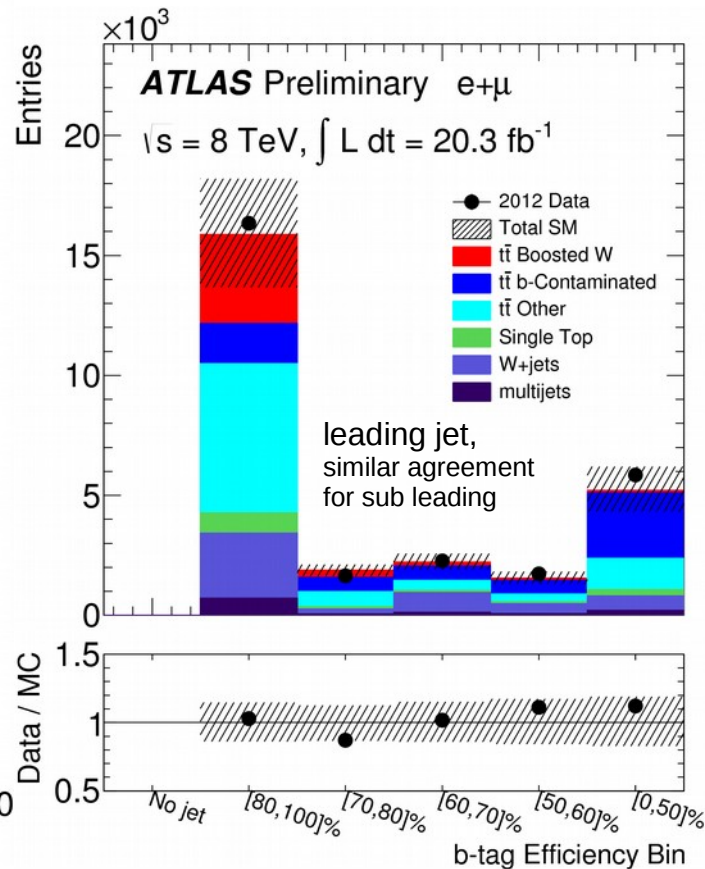
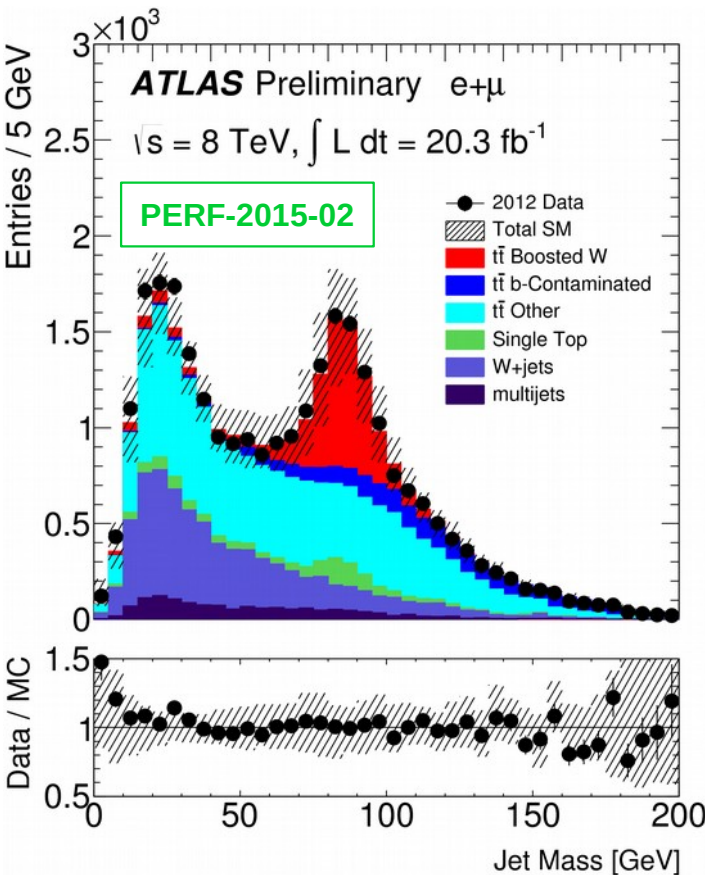
## Semi-leptonic $t\bar{t}$ selection of hadronically decaying $W$

MC  $t\bar{t}$  divided into:  $t\bar{t}$  Boosted  $W$ ,  $t\bar{t}$  b-contaminated,  $t\bar{t}$  others (using MC matching)

W+jets and multi-jets QCD from data-driven methods

Others from simulation

## Tests of the 3 different input variables in data → well modelled





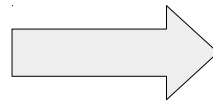
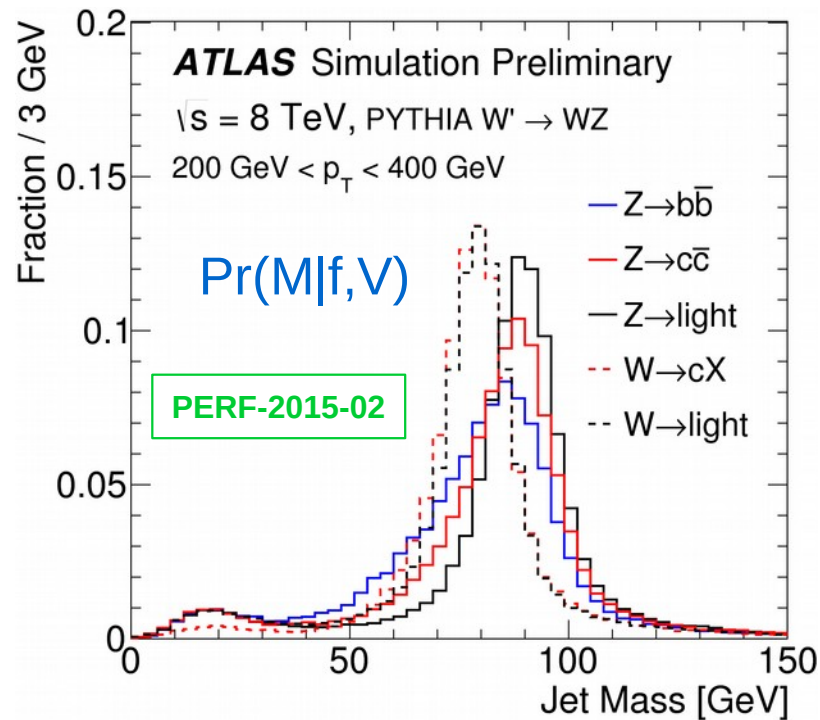
# Tagger definition

Likelihood tagger: templates build for each possible boson decay channel

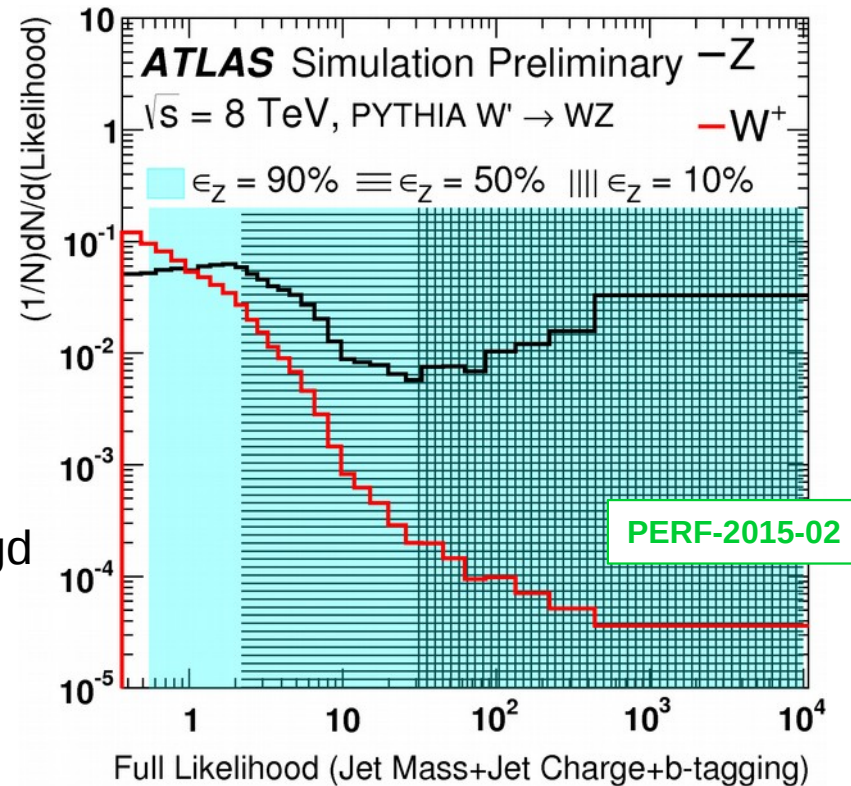
$$\Pr(M, Q, B|V) = \sum_f \Pr(f|V) \Pr(M|f, V) \Pr(Q|f, V) \Pr(B|f, V) \quad \begin{array}{l} f: \{bb, cc, cs, cd, \text{light}\} \\ V: \{W, Z\} \end{array}$$

with  $p(B|f, V) = p(B_{\text{lead}}|f, V)p(B_{\text{sub-lead}}|f, V)$

This factorisation based on the boson decay channel is very good.  
The tagger is then build from likelihood ratio.



1D histo build from the ratios signal/bkgd of each  $\Pr(M, Q, B|V)$





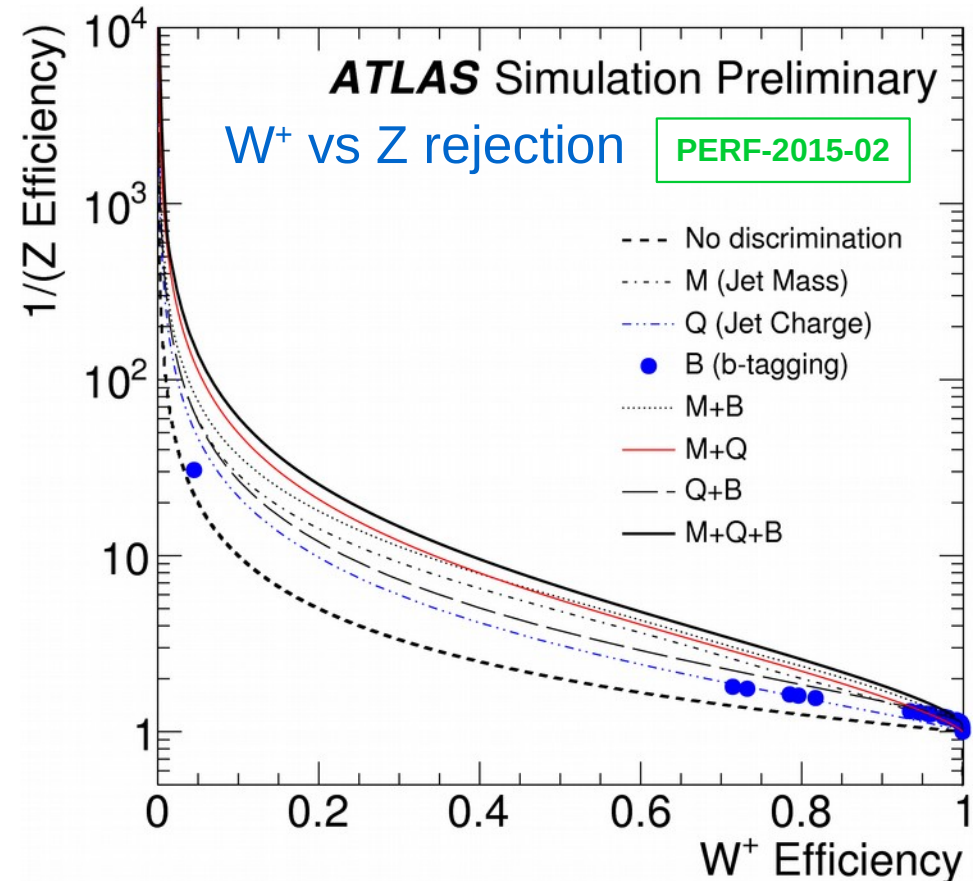
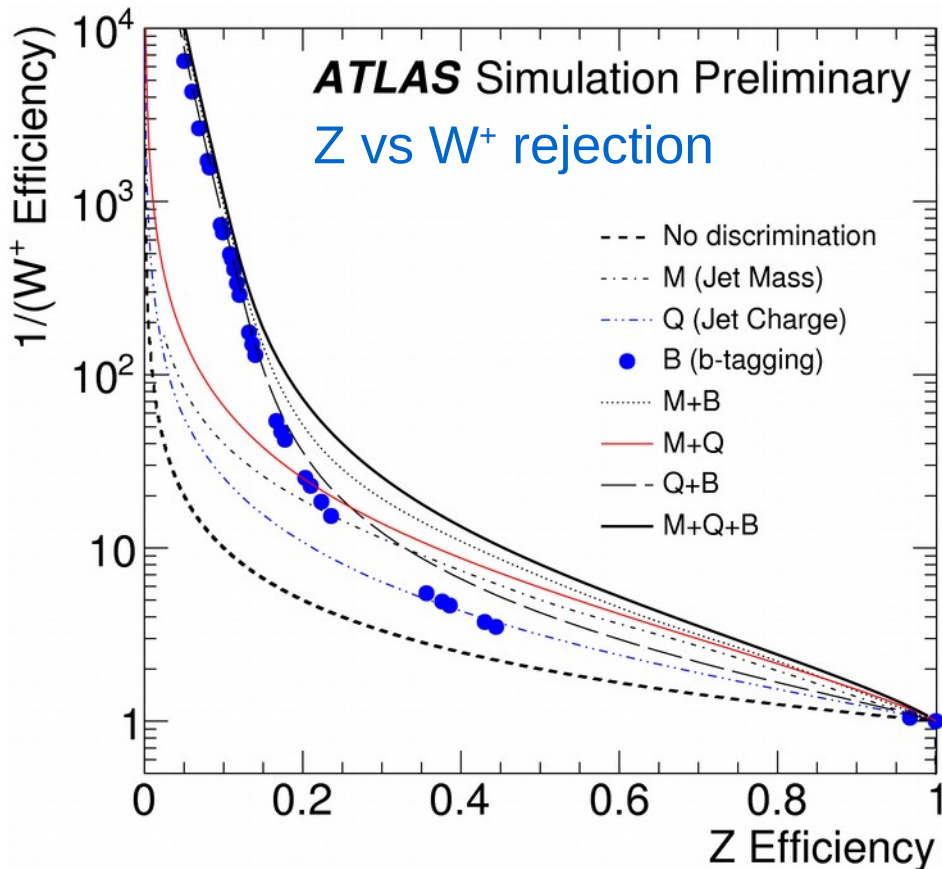
# Tagger efficiencies

Efficiencies for Z vs W<sup>+</sup> rejection and for W<sup>+</sup> vs Z rejection

Example of application:  $(50\% \times \sigma(WZ)) / (12\% \times \sigma(WW)) = 50/12 \times 20\% = 83\%$

For Z-tagging, at low eff.: big gain from b-tagging because of  $Z \rightarrow b\bar{b}$

No syst. uncertainties (next slide)





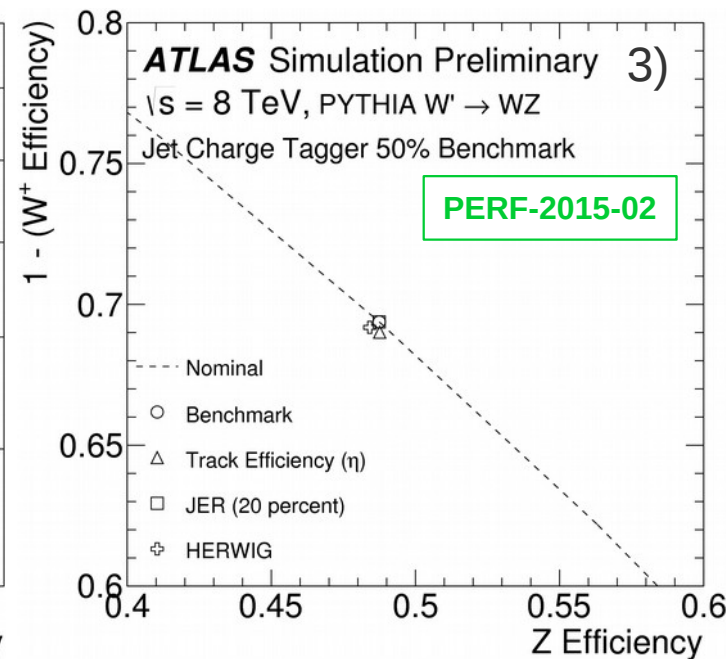
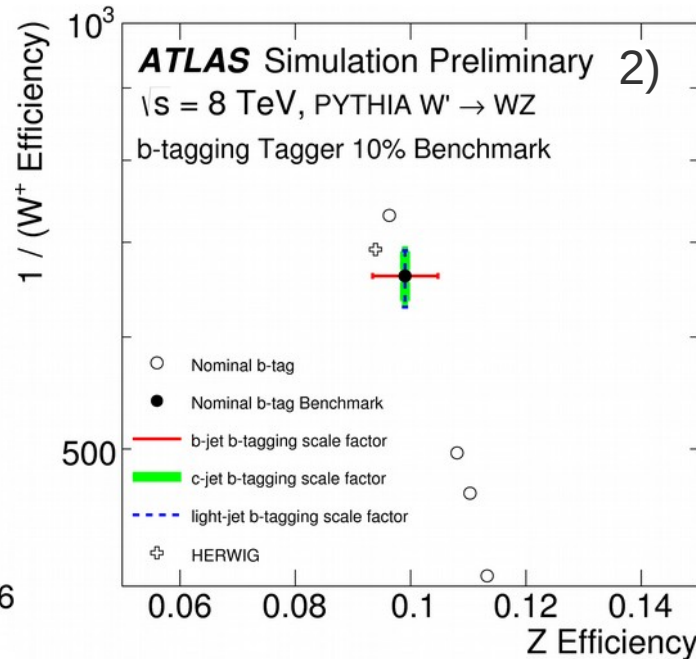
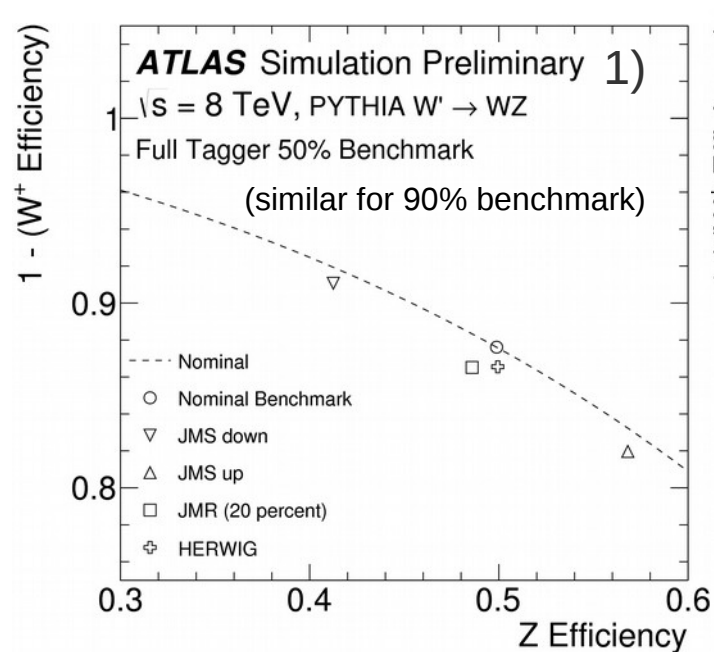


# Systematic uncertainties

## Sources of systematic uncertainties:

- 1) Calibration of large-R and small-R jets: JMS moves on the ROC curve  
JMR reduces performances
- 2) b-tagging uncertainties: from MC/Data discrepancies  
relevant at low Z-tagger eff.,  
b-tag affecting Z, c- and light-tag affecting W
- 3) modelling of track reconstruction: small impact in this  $p_T$  range

## Dominant uncertainties: jet mass scale and resolution





# Conclusion



## Many grooming techniques and variables have been tested in Run1:

- grooming: Background rejection in mass window  
Pile-Up stability  
→ **anti-kt R=1.0 trimmed  $f_{\text{cut}}=5\%$ ,  $R_{\text{sub}}=0.2$**
- variables: Best performers:  $\tau_{21}^{\text{wta}}$ ,  $C_2^{(\beta=1)}$ ,  $D_2^{(\beta=1)}$

**W-tagger in Run1, 2 working points (medium and tight)**

## Early W-tagger and Z-tagger for Run2:

Based on Run1 study, subset tested

→ **anti-kt R=1.0 trimmed  $f_{\text{cut}}=5\%$ ,  $R_{\text{sub}}=0.2 + D_2^{(\beta=1)}$**

Uncertainties from Run1 + Run2 simulation variations

**W-tagger and Z-tagger in Run2, 2 working points (medium and tight)**

## W and Z separation in Run1:

Based on 3 discriminating variables: mass, charge and b-tagging

**Z selection tagger, W selection tagger provided**

Calibration of the tagger in the data in Run2



# Backup slides

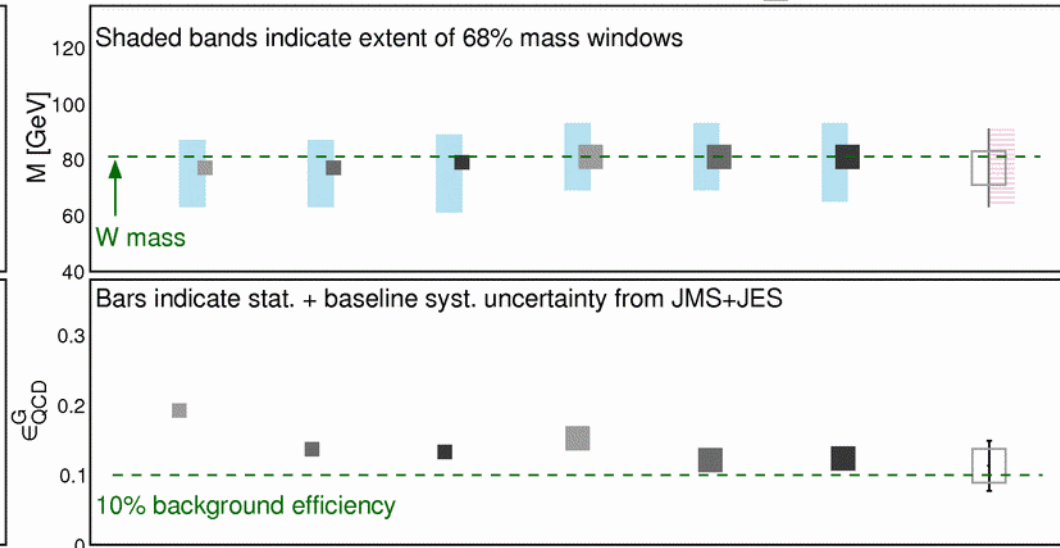
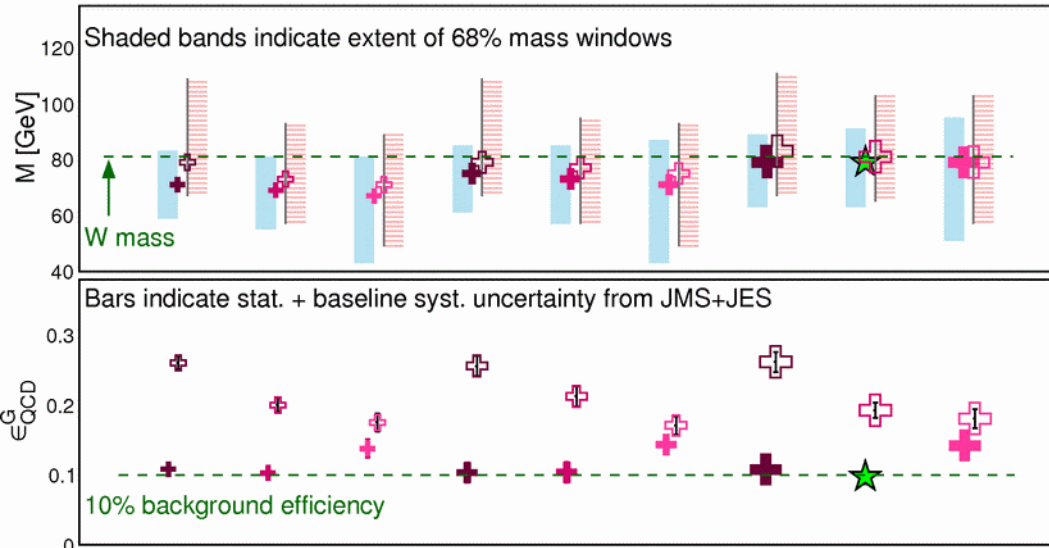


# W-tagging in Run1

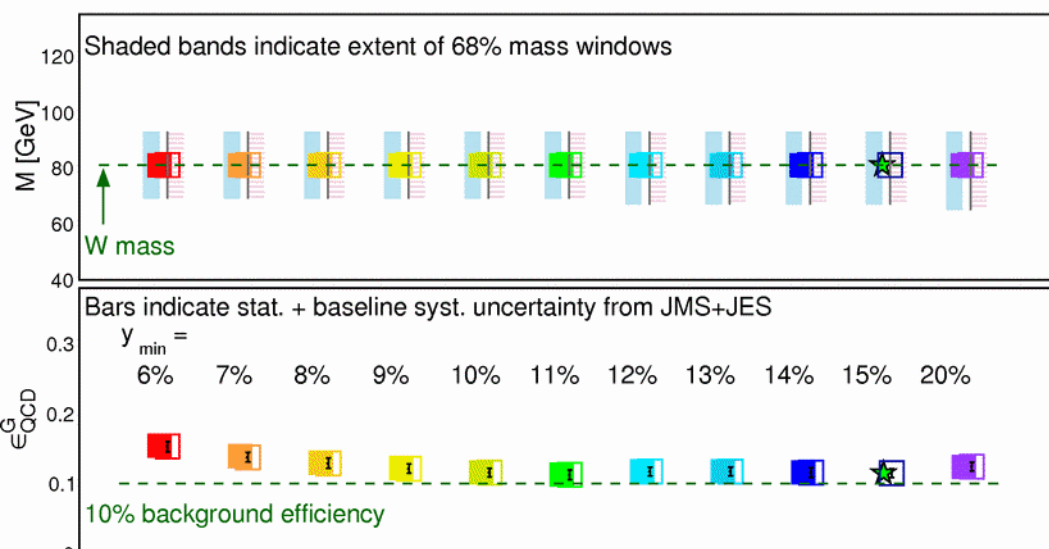
PERF-2015-03



Pruned jets,  $R=1.0$   $350 < p_T^{\text{Truth}} < 500 \text{ GeV}$   $|\eta^{\text{Truth}}| < 1.2$   $C/A$ -prun.  $k_t$ -prun.  $1/8$   $10\%$   $1/4$   $Z_{\text{cut}} = 15\%$   $20\%$   $\sqrt{s} = 8 \text{ TeV}$  **ATLAS Preliminary**  $\mu_{\text{max}} = 1$   $\mu_{\text{max}} = 2/3$   $R_{\text{sub}} = 0.3$   $y_{\text{min}} = 9\%$   $20\%$  **ATLAS Preliminary**  $\text{Pythia8 simulation}$   $\sqrt{s} = 8 \text{ TeV}$



Split-filtered jets,  $R=1.2$   $350 < p_T^{\text{Truth}} < 500 \text{ GeV}$   $|\eta^{\text{Truth}}| < 1.2$   $\mu_{\text{max}} = 1$   $\mu_{\text{max}} = 2/3$   $R_{\text{sub}} = 0.3$  **ATLAS Preliminary**  $\text{Pythia8 simulation}$   $\sqrt{s} = 8 \text{ TeV}$

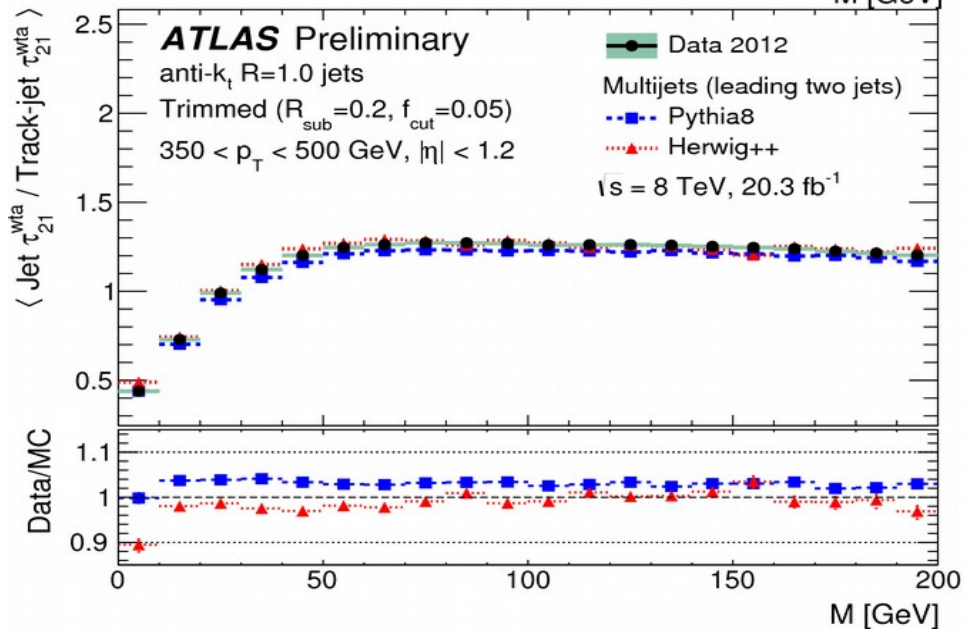
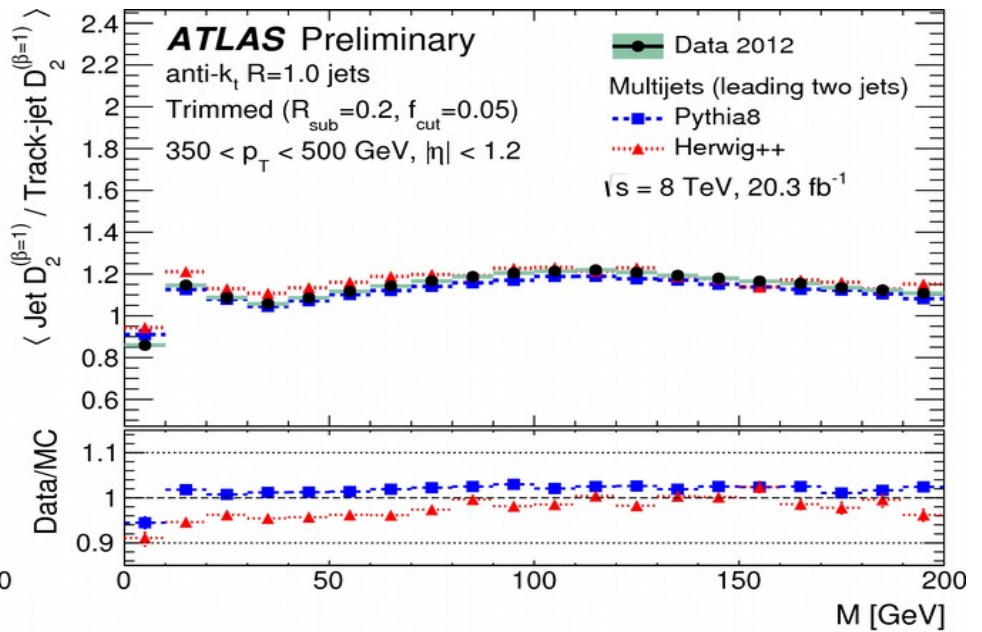
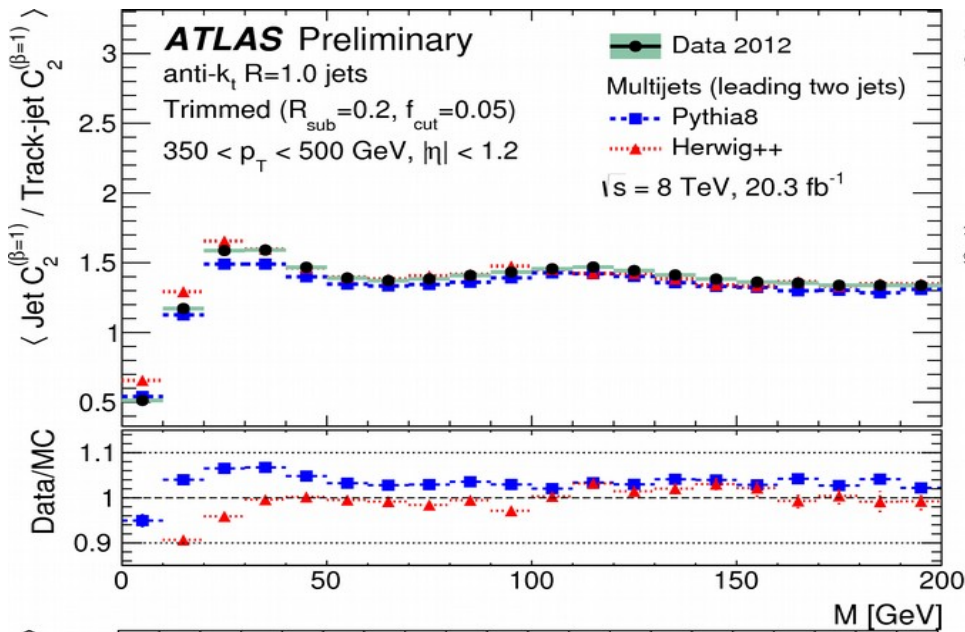


Performance of different grooming techniques



# W-tagging in Run1

PERF-2015-03



Track-jet double-ratio for  $C_2$ ,  $D_2$ ,  $\tau_{21}^{\text{wta}}$

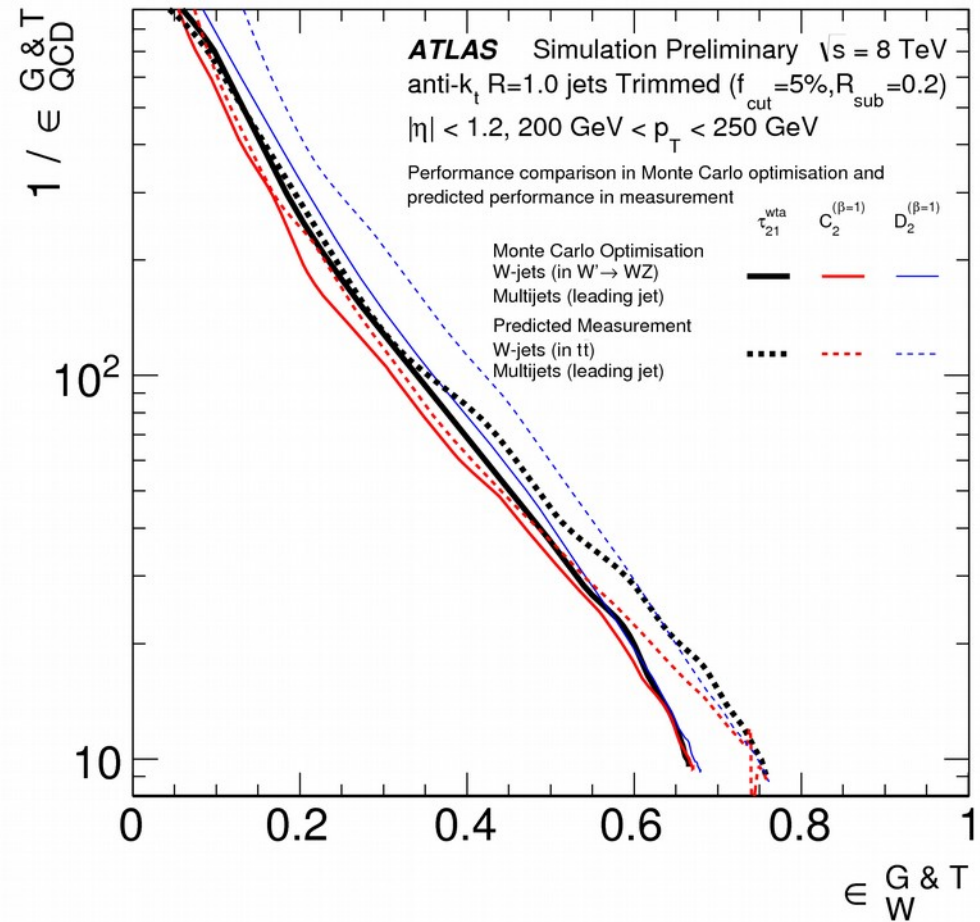
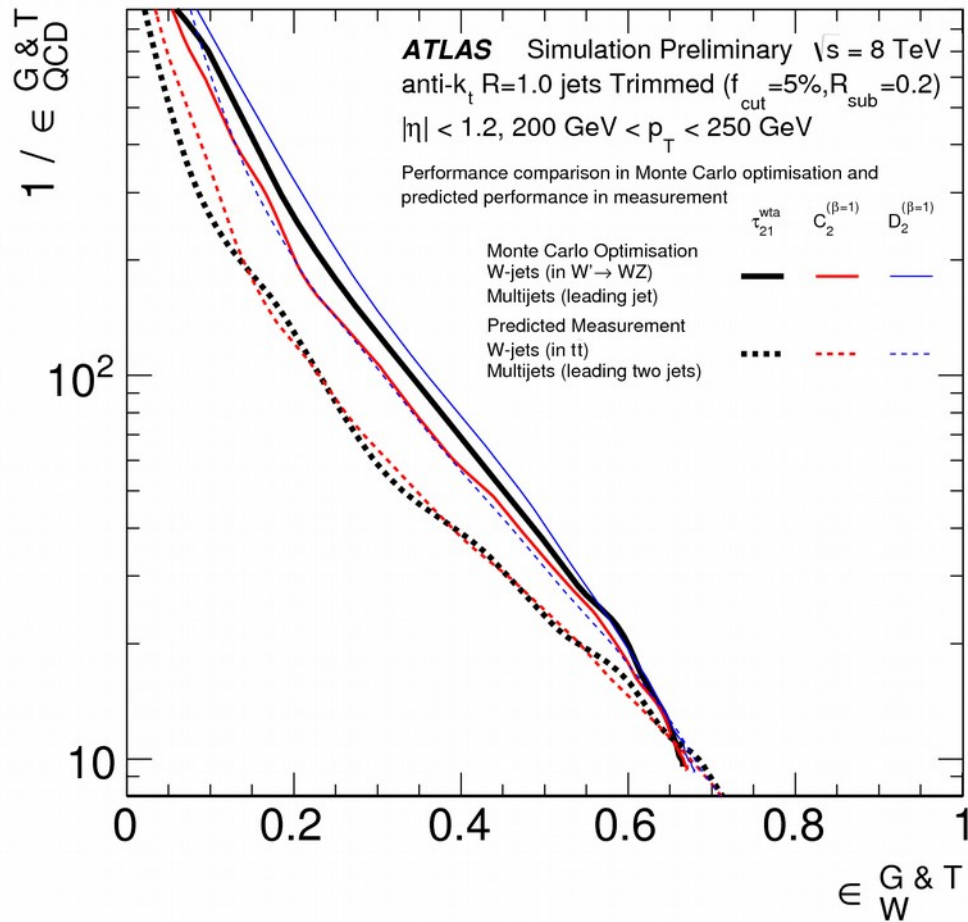


# W-tagging in Run1

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JG|U

## Differences between W-jets in $W'$ simulations and W-jets in $t\bar{t}$ predicted measurement



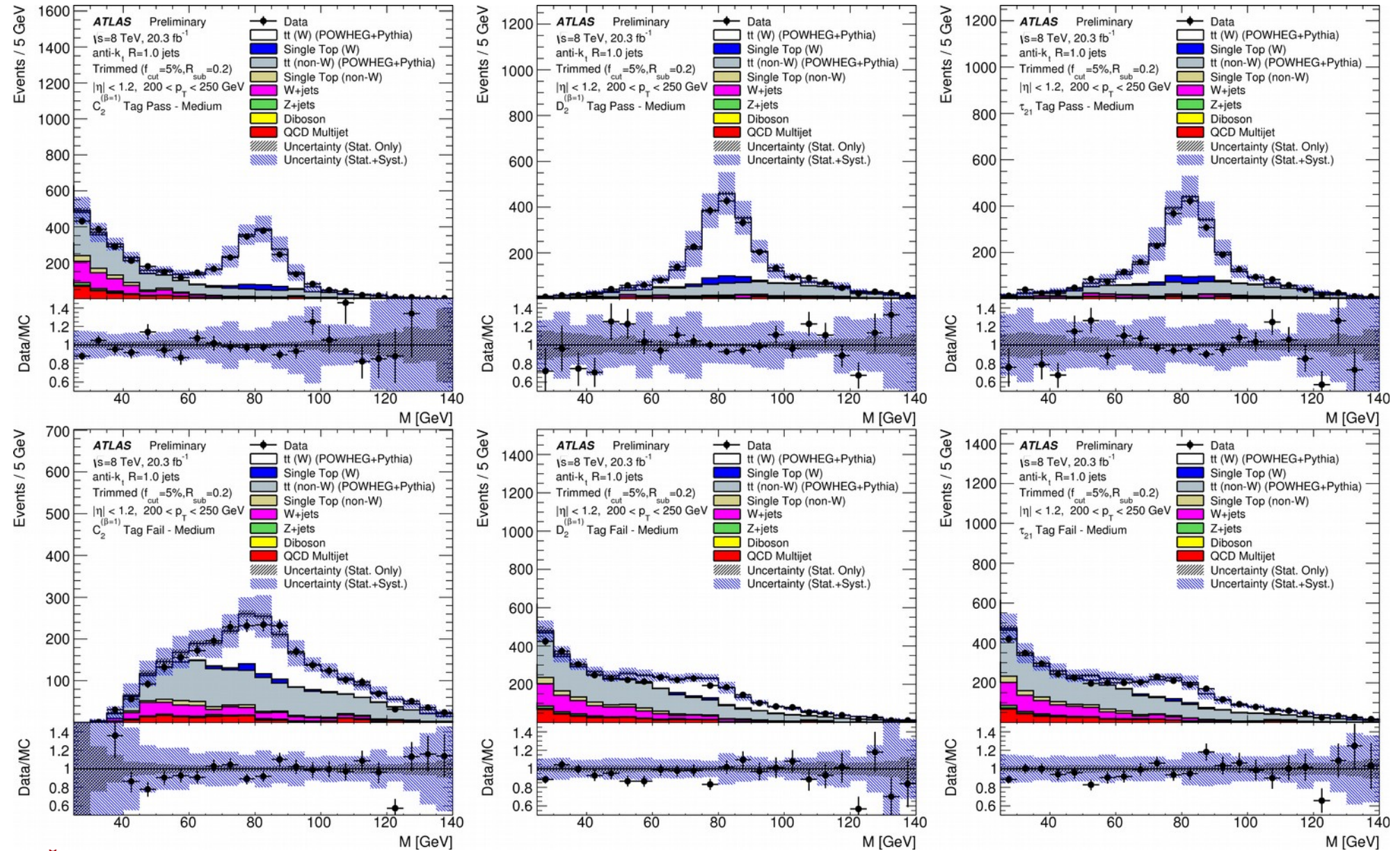


# W-tagging in Run1

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Mass distribution for passing (top) and failing (bottom) medium WP





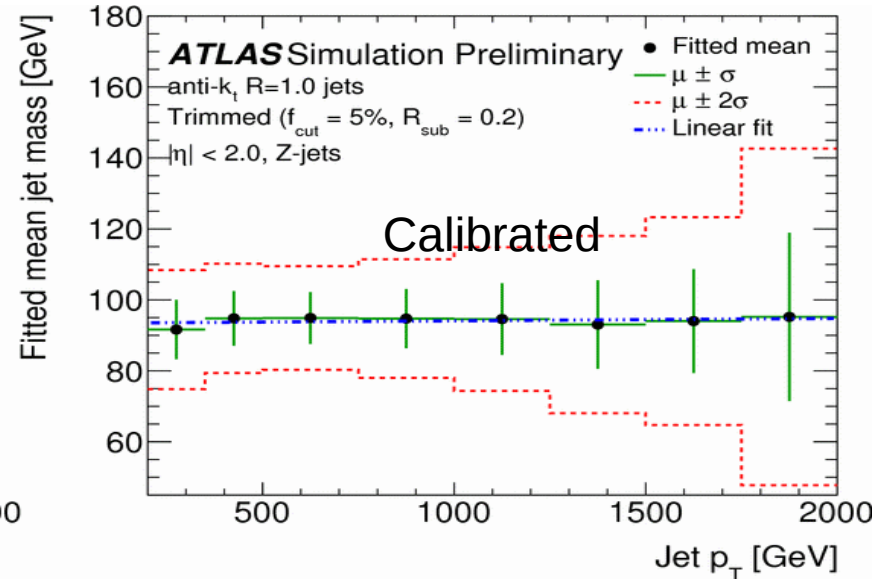
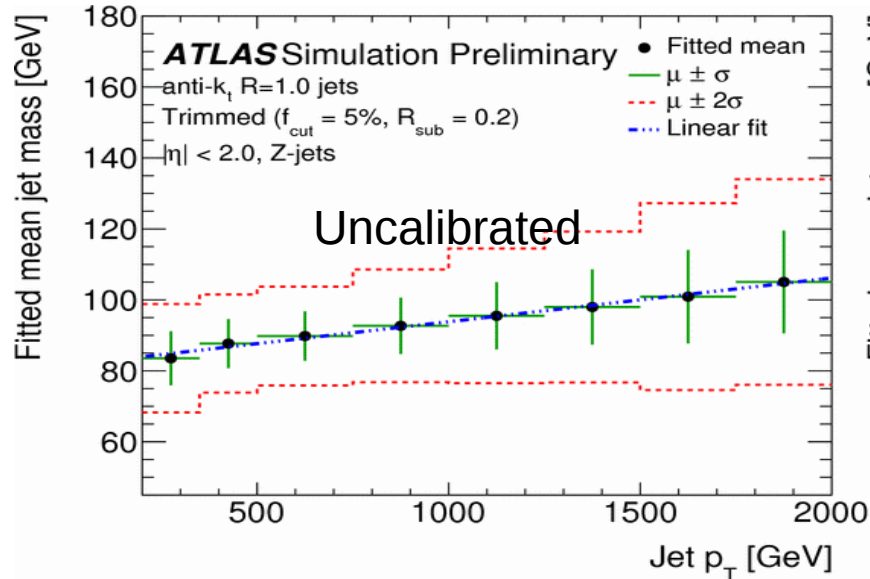


# W-tagging in Run2

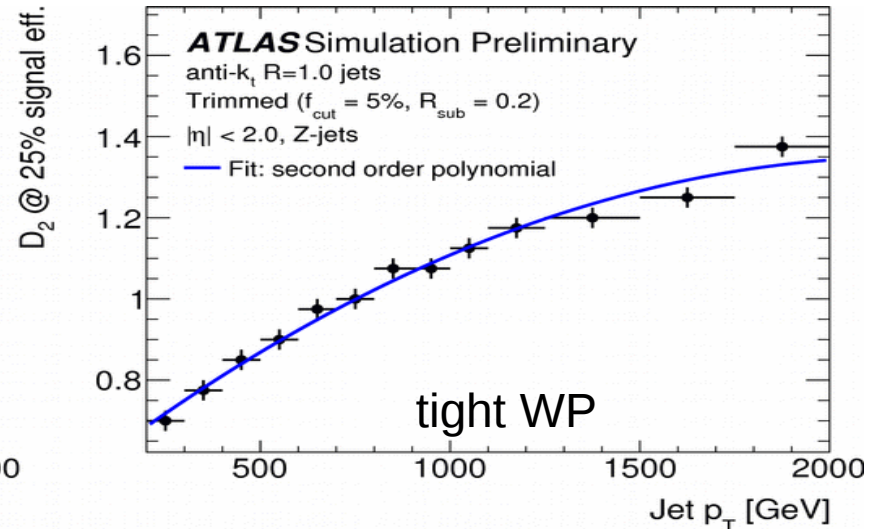
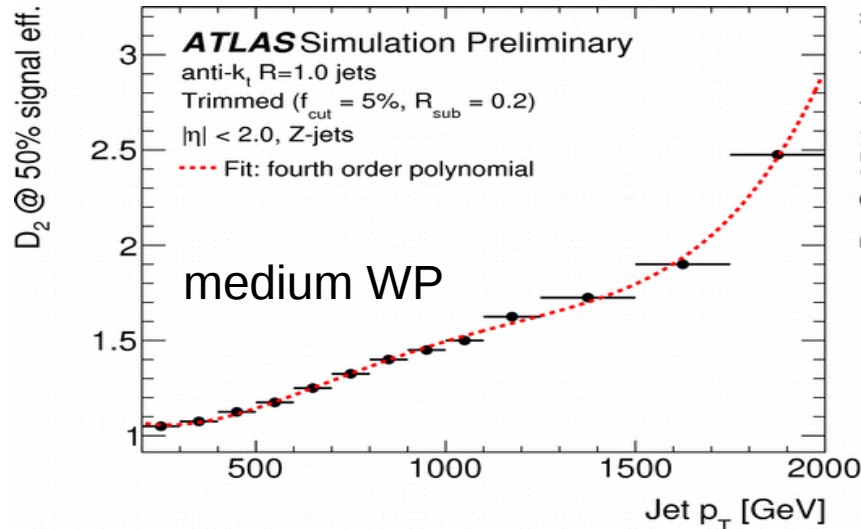
ATL-PHYS-PUB-2015-033



## Mass calibration: (here for Z-boson)



## Cut parametrization: (here for Z-boson)



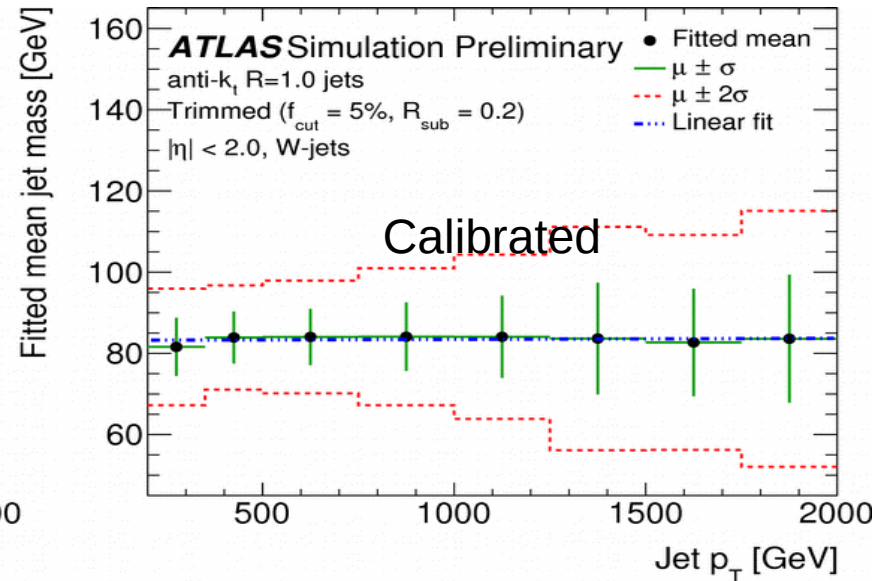
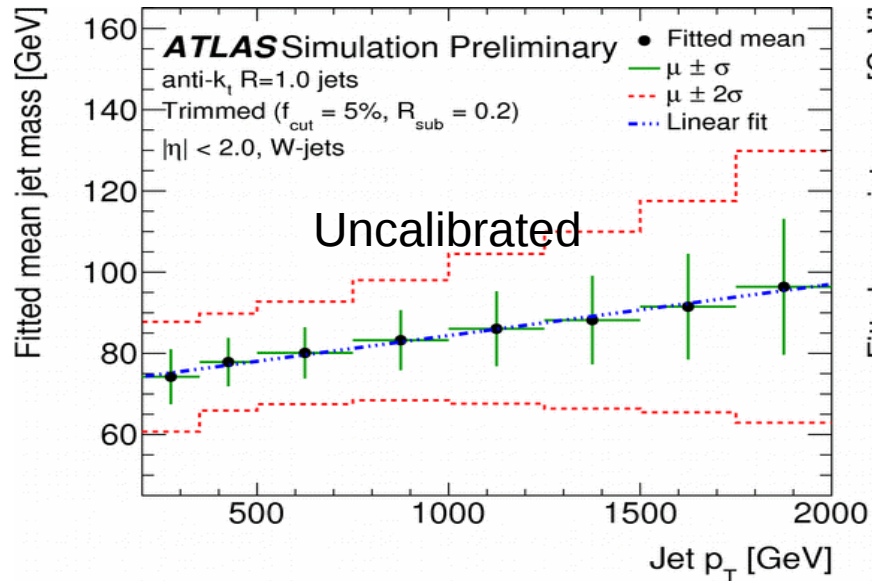


# W-tagging in Run2

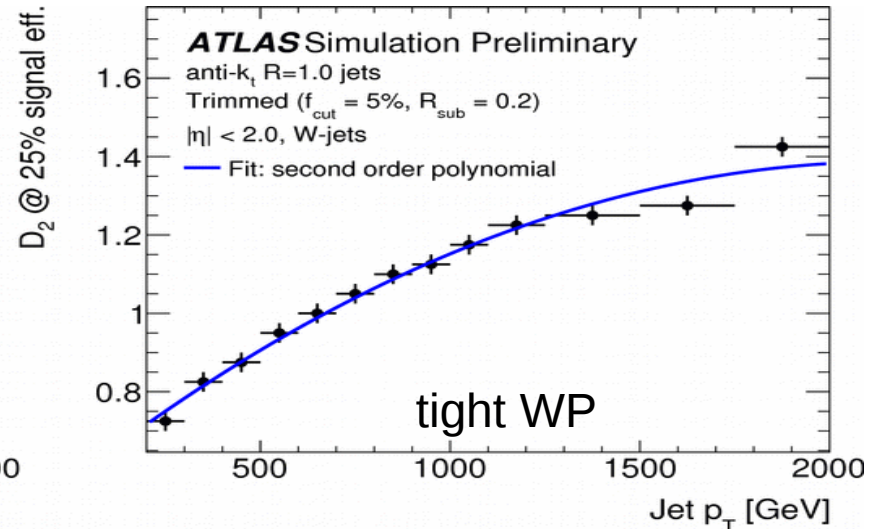
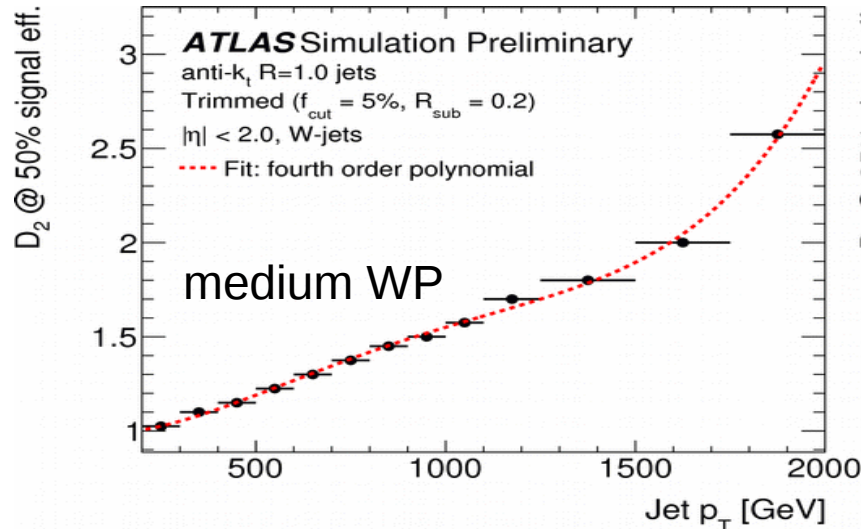
ATL-PHYS-PUB-2015-033



## Mass calibration: (here for W-boson)



## Cut parametrization: (here for W-boson)





# W-tagging in Run2

ATL-PHYS-PUB-2015-033

JG|U

