

b-tagging calibration using the $t\bar{t}$ PDF method with the ATLAS experiment in Run II

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Special thanks to the whole $t\bar{t}$ PDF *b*-calibration team

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b-tagging calibration in ATLAS

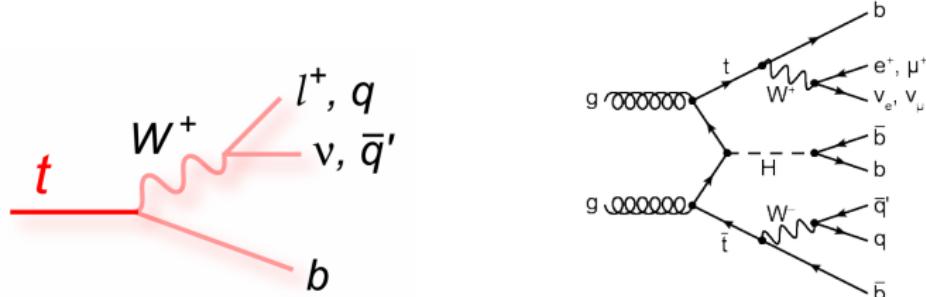
b-tagging calibration

Motivation of flavour-tagging



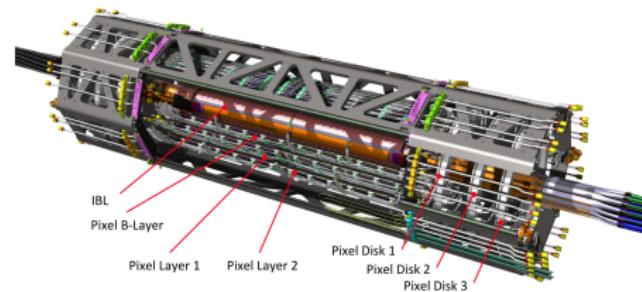
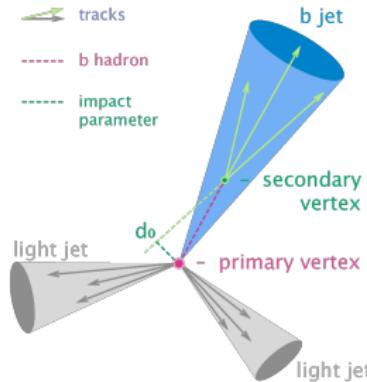
Why is *b*/flavour-tagging so important?

- LHC collides protons \Rightarrow always have many jets (also from pileup)
- LHC is a “top factory” $\Rightarrow \text{BR}(t \rightarrow Wb) \approx 100\%$
- *b*-tagging one of leading uncertainties in many ATLAS top analyses
- $\text{BR}(H \rightarrow b\bar{b}) \approx 58\%$
 \Rightarrow *b*-tagging also important for $t\bar{t}H$, VH & BSM analyses
- Some analyses must reject *b*-jets! (e.g. $H \rightarrow \tau\tau$)



Multivariate analysis approach to identify b -jets

- Most vital input: charged particle tracks reconstructed in inner detector
- Information from tracks associated to jets used in different algorithms
- Run II: Insertable B-Layer (IBL)! → positive impact on b -tagging
(b -tagging performance plots with/without IBL)
- Tracking improved in Run II (+), but jets have higher boost on average (-)

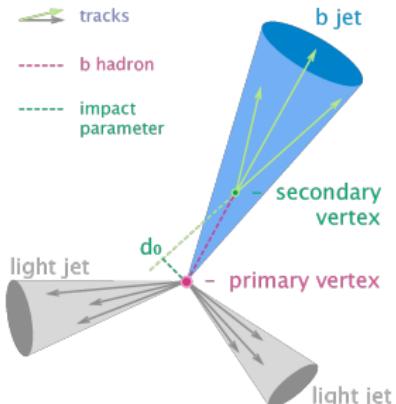


b-tagging calibration

The MV2c10 algorithm in ATLAS

ATLAS uses 3 distinct basic *b*-tagging algorithms

1. Impact parameter based
2. Inclusive secondary vertex reconstruction
3. Decay chain multi-vertex reconstruction



- Complementary algorithms, output combined in multivariate discriminant
 - Trained on $\approx 5M$ $t\bar{t}$ events:
 - Signal: *b*-jets
 - Background: mixture of 10% c-jets + 90% light-jets
- \Rightarrow Best separation for different jet flavours: **MV2c10**

Improvements from Run I to Run II: ATL-PHYS-PUB-2015-022
MV2 optimisation studies: ATL-PHYS-PUB-2016-012

b-tagging calibration

Why do we need calibration?



What does 'calibration' mean in *b*-tagging?

- Choose working point of your algorithm, e.g. $\epsilon_b = 70\%$
 \Rightarrow 70% of *b*-jets are *b*-tagged in MC
 \Rightarrow What about data?
- Perform calibration: Estimate ϵ_b^{data} and corresponding uncertainties
- Analysers apply SF $\kappa = \epsilon_b^{\text{data}} / \epsilon_b^{\text{MC}}$ to correct data/MC difference
- κ should be close to 1 \rightarrow shows good performance of algorithm

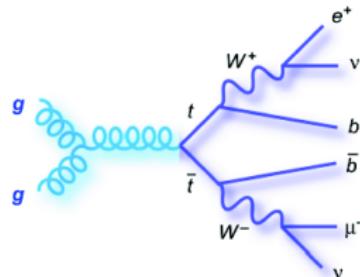
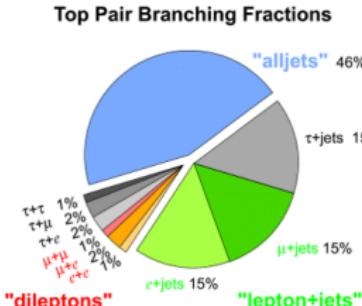
Examples of different strategies & environments

- Run II: *b*-jets in dileptonic $t\bar{t}$ with combinatorial likelihood approach (Run I: ATLAS-CONF-2014-004)
- *b*-jets in dileptonic $t\bar{t}$ with tag-and-probe method
- *b*-jets in l+jets $t\bar{t}$ with tag-and-probe method
- light-flavour jets in di-jet events with negative tag method
- *c*-jets with "JetFitterCharm" in l+jets $t\bar{t}$ (ATL-PHYS-PUB-2015-001)

The Probability Density Function likelihood method in dileptonic $t\bar{t}$ events

b -tagging calibration in dileptonic $t\bar{t}$ using PDF likelihood method

- Select as many events rich in b -jets as possible
- Dileptonic $t\bar{t}$ events \rightarrow pure final state ($l^+ l^- \nu \bar{\nu} b\bar{b}$), low c -jet contribution
 \Rightarrow Main background: light-jets from initial/final state radiation
- Split into 6 channels for control: ($ee, \mu\mu, e\mu$) \times (2j, 3j)
- Perform fit to likelihood function in all data events
 \Rightarrow Extract ϵ_b as function of jet p_T
- Derive data/MC scale factors and systematic uncertainties



Per-event likelihood function for two jet case:

$$\begin{aligned}\mathcal{L}(p_{T,1}, p_{T,2}, w_1, w_2) = & [f_{bb} \mathcal{P}_{bb}(p_{T,1} p_{T,2}) \mathcal{P}_b(w_1|p_{T,1}) \mathcal{P}_b(w_2|p_{T,2}) \\ & + f_{bl} \mathcal{P}_{bl}(p_{T,1} p_{T,2}) \mathcal{P}_b(w_1|p_{T,1}) \mathcal{P}_l(w_2|p_{T,2}) \\ & + f_{ll} \mathcal{P}_{ll}(p_{T,1} p_{T,2}) \mathcal{P}_l(w_1|p_{T,1}) \mathcal{P}_l(w_2|p_{T,2}) \\ & + 1 \leftrightarrow 2] / 2\end{aligned}$$

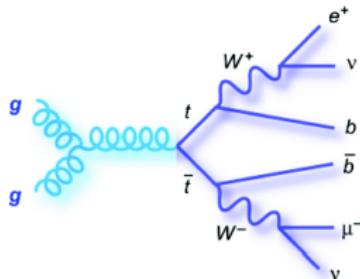
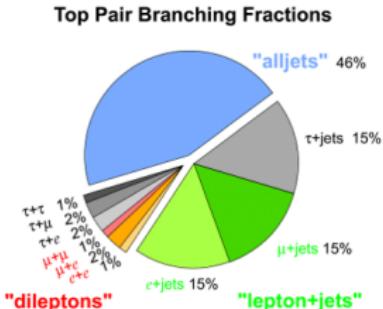
- $f_{bb}, f_{bl}, f_{ll} = 1 - f_{bb} - f_{bl}$: two jet flavour fractions
- $\mathcal{P}_f(w|p_T)$: PDF for b -tag weight for jet flavour f , depending on jet p_T
- $\mathcal{P}_{f_1 f_2}(p_{T,1}, p_{T,2})$: two-dim PDF for combination of $[p_{T,1}, p_{T,2}]$ & $[f_1, f_2]$
- PDFs implemented as binned histograms

Define total likelihood function, to be maximised on all data events:

$$\mathcal{L}_{\text{tot}} = \exp(-N) \prod_{i=1}^{N_{\text{events}}} \mathcal{L}(p_{T,1,i}, p_{T,2,i}, w_{1,i}, w_{2,i})$$

Extract $\mathcal{P}_b(w|p_T)$ from fit $\Rightarrow \underline{\epsilon_b(p_T)} = \int_{w_{\text{cut}}}^{\infty} dw' \mathcal{P}_b(w'|p_T)$

- Exactly 2 oppositely charged (OS) leptons (e, μ):
 - Trigger-matched, tight isolation, no overlap with jets
 - $p_T > 28 \text{ GeV}$, $|\eta_\mu| < 2.5$, $|\eta_{\text{el}}| < 2.47$ (exclude $1.37 < |\eta_{\text{el}}| < 1.52$),
- Exactly two or three jets:
 - Anti- k_t with $\Delta R = 0.4$ from the calorimeter
 - jet $p_T \geq 20 \text{ GeV}$, jet $|\eta| < 2.5$
- End up with 6 channels:
 - $ee, \mu\mu$ dominated by $Z + \text{jets}$ (smaller background: single top, Diboson)
 \Rightarrow Allow to constrain $Z + \text{jets}$ and improve stat
 - $e\mu$ highest purity; main bkgd: single top \Rightarrow Consider $e\mu + 2j$ for calibration



- Analyse ATLAS 2015+2016 data at $\sqrt{s} = 13$ TeV with $\int L dt = 36.1 \text{ fb}^{-1}$
- Compare to MC simulation for dileptonic $t\bar{t}$ signal & background:

Process	Nominal MC	Alternative MC
Signal: $t\bar{t}$	Powheg+Pythia8	Sherpa 2.2.1
Background: $Z + \text{jets}$	Sherpa 2.2.1	Powheg+Pythia8
Single top	Powheg+Pythia6	-
Diboson	Powheg+Pythia8	Sherpa 2.2.1
Lepton fakes	From data	-
$W + \text{jets}$	Sherpa 2.2.1	-

This is work in progress, more alternative samples are being produced!

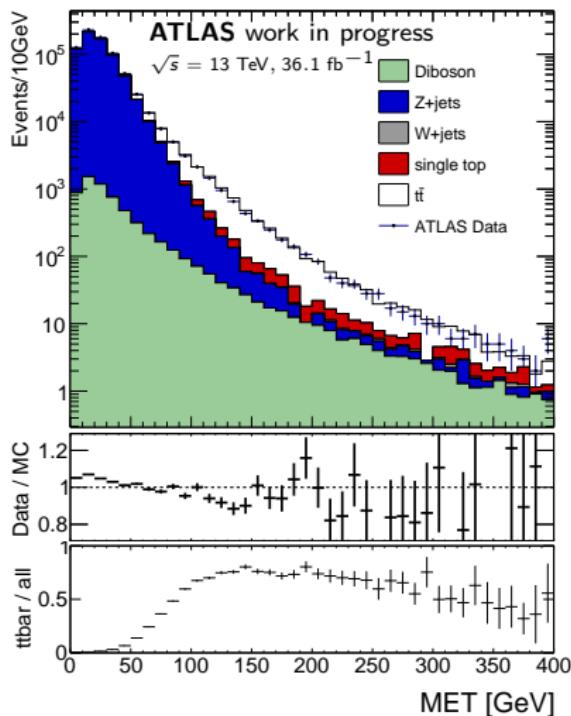
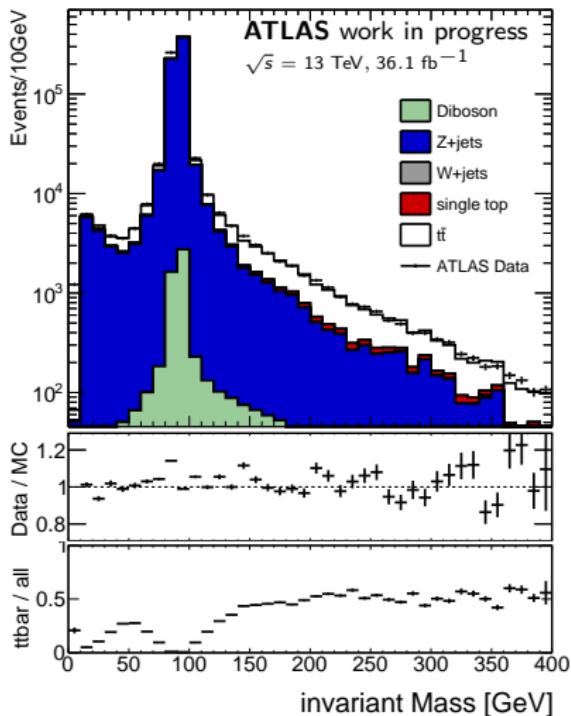
PDF Likelihood method

Initial control plots in ee+2j



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Z+jets dominant background in ee, $\mu\mu$



⇒ Require $E_T^{\text{miss}} > 60 \text{ GeV}$ and $50 \text{ GeV} < m_{ll} < 80 \text{ GeV}$ or $m_{ll} > 100 \text{ GeV}$

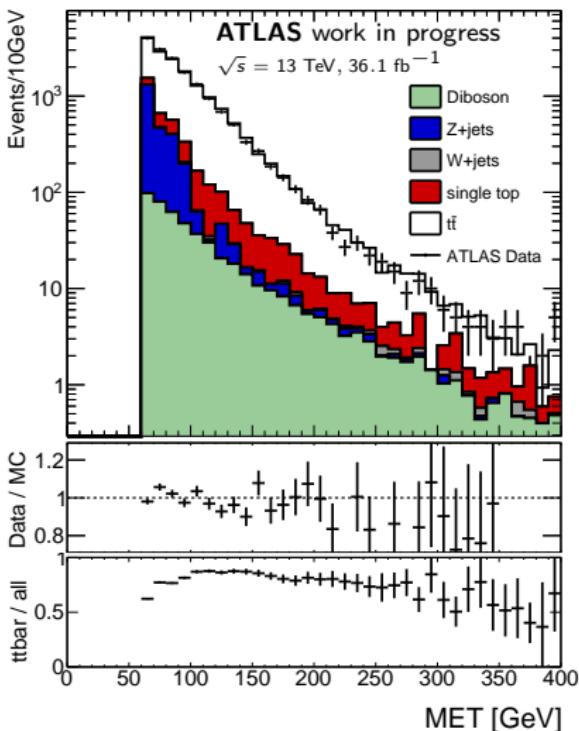
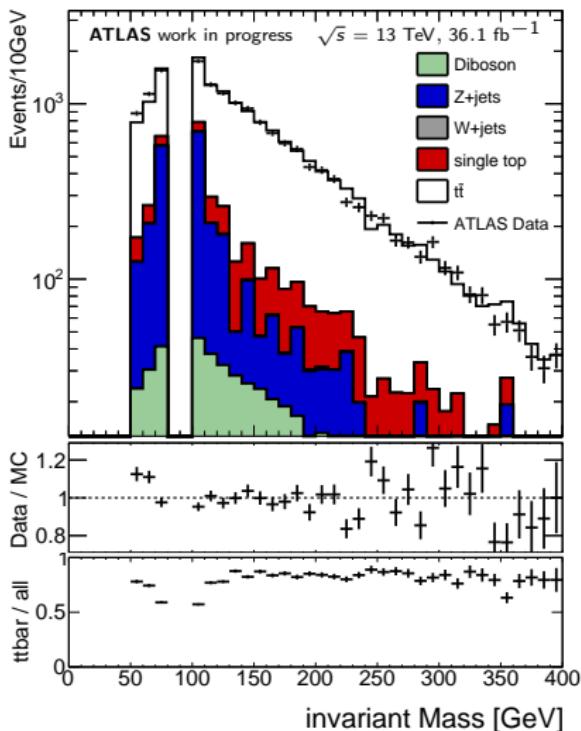
PDF Likelihood method

Apply cut in ee+2j



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After cut on $E_T^{\text{miss}} > 60 \text{ GeV}$ and $50 \text{ GeV} < m_{\parallel} < 80 \text{ GeV}$ or $m_{\parallel} > 100 \text{ GeV}$:



PDF Likelihood method

Additional cut in $e\mu$



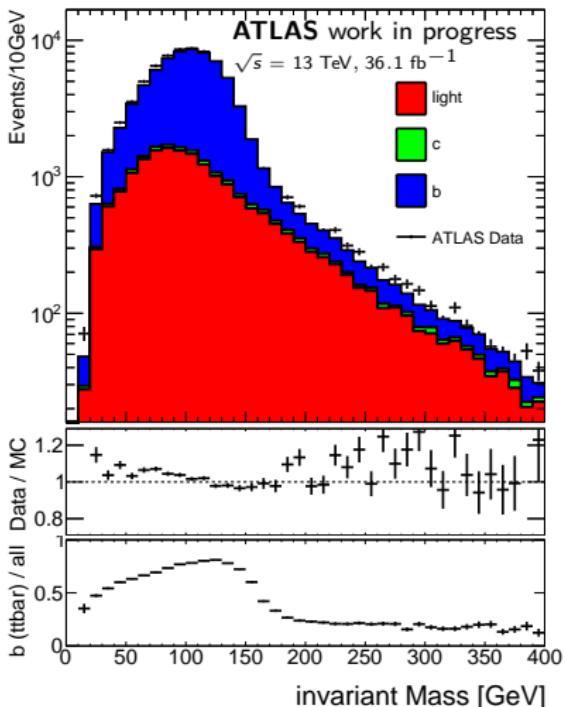
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- $e\mu$ channel has highest purity ($\approx 7 - 14\%$ larger than $ee, \mu\mu$)
- Low $Z + \text{jets}$ background
 \Rightarrow no MET or m_{\parallel} cuts applied

Think of further cut:

- Build all possible pairs of leptons and jets \Rightarrow look at $m_{l,j}$
- If lepton and jet from same top, $m_{l,j}$ has end point at $\approx m_{\text{top}}$
- Combine leptons with jets such that $(m_{l_1,j_1}^2 + m_{l_2,j_2}^2)$ minimal
 \Rightarrow Require $m_{l,j} < 175$ GeV for one or both pairs

Check $m_{l,j}$ split by flavours:



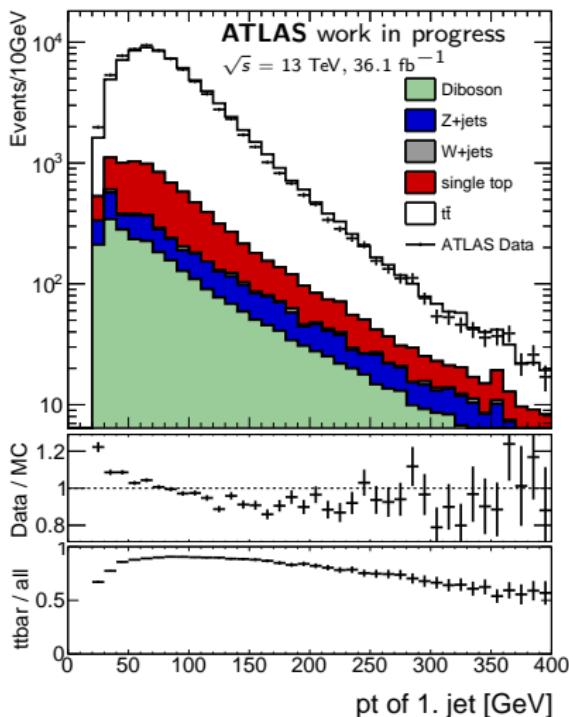
PDF Likelihood method

Apply 2 lepton cut in $e\mu + 2j$

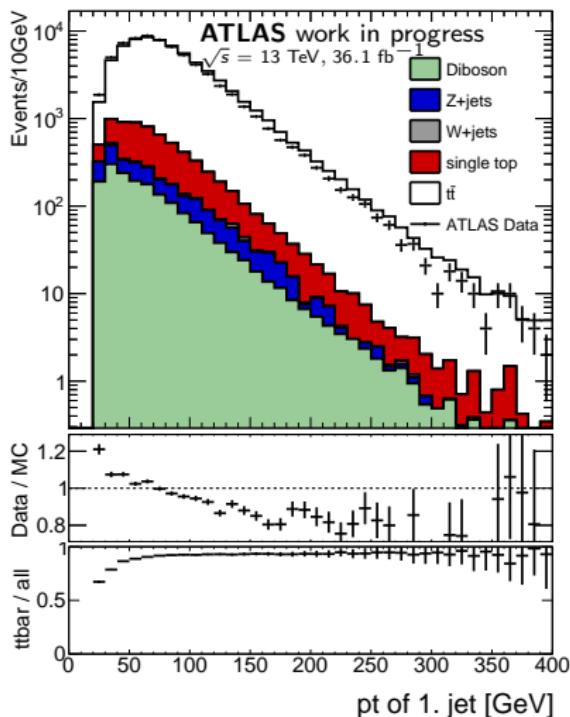


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Before cut



After cut



⇒ Signal purity increases especially in high jet p_T regions

PDF Likelihood method

Renormalisation of MC



Select regions to extract data/MC renormalisation scale factors for $t\bar{t}$ & $Z + \text{jets}$
⇒ Improve Data/MC agreement before fit
Use same-sign (SS) lepton events to estimate non-prompt and fake leptons

Process	$e\mu$	$ee, \mu\mu$
$t\bar{t}$	2 OS leptons Different lep flavour	2 OS leptons Same lep flavour $50 \text{ GeV} < m_{ll} < 80 \text{ GeV}$ or $m_{ll} > 100 \text{ GeV}; E_T^{\text{miss}} > 60 \text{ GeV}$
$Z + \text{jets}$	2 OS leptons Same lep flavour $80 \text{ GeV} < m_{ll} < 100 \text{ GeV}$	2 OS leptons Same lep flavour $80 \text{ GeV} < m_{ll} < 100 \text{ GeV};$ $E_T^{\text{miss}} > 60 \text{ GeV}$
Non-prompt/ fake leptons	2 SS leptons Different lep flavour	2 SS leptons Same lep flavour $50 \text{ GeV} < m_{ll} < 80 \text{ GeV}$ or $m_{ll} > 100 \text{ GeV}; E_T^{\text{miss}} > 60 \text{ GeV}$

- $Z + \text{jets}$ SF between 0.96 - 1.05
- $t\bar{t}$ signal SF between 0.98 - 1.03
- Contribution from fakes found to be negligible

Main systematics:

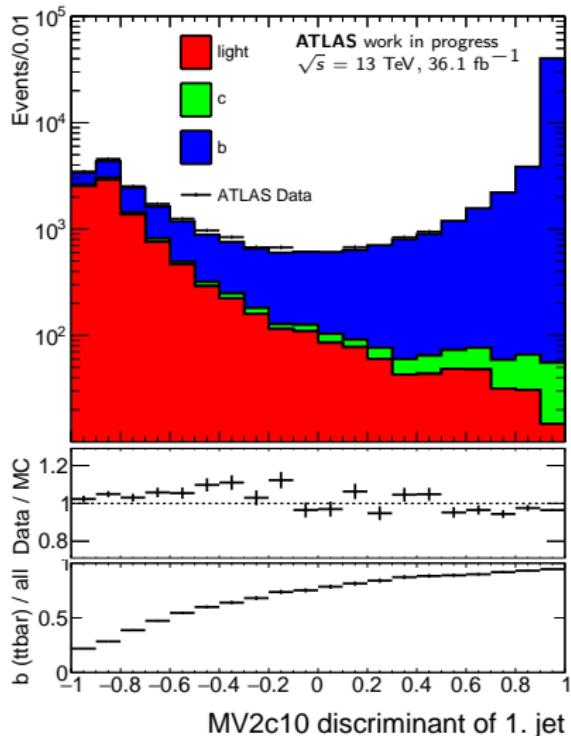
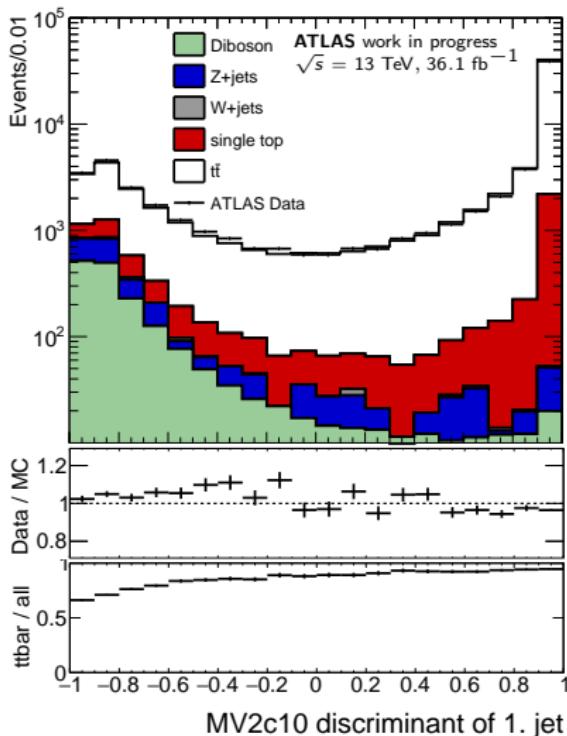
- Monte Carlo Generator
 - ⇒ Currently: only Powheg+Pythia8 vs. Sherpa available
 - ⇒ Future: Powheg vs. MadGraph5_aMC@NLO; Pythia8 vs. Herwig7
- Pileup modelling
- Jets
 - Jet energy scale
 - Jet energy resolution
 - Jet vertex tagger
- Leptons
 - Trigger
 - Identification
 - Isolation

Fit input

MV2c10 for first jet in $e\mu+2j$



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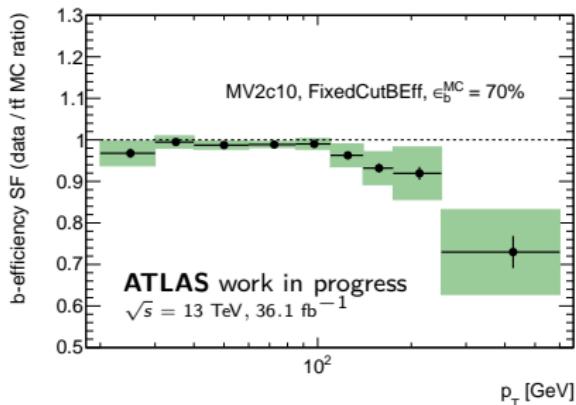


Fit result

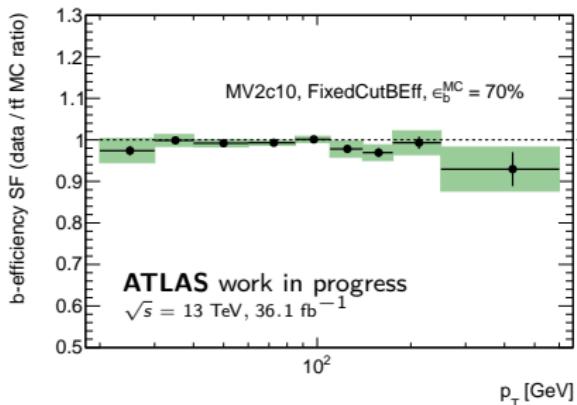
Data/MC SF for MV2c10 (Fixed cut)



$\epsilon_b = 70\% : \text{Before cut}$



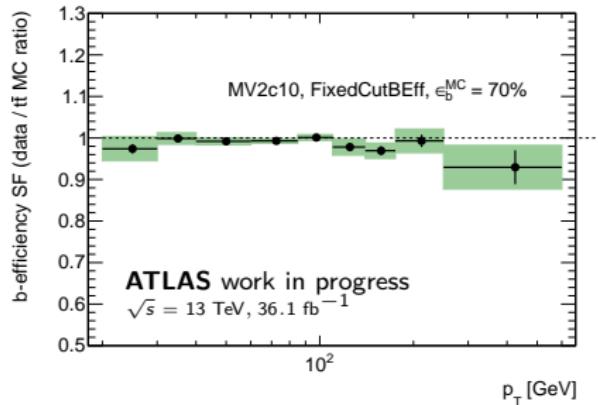
After 2 lepton cut



- Scale factor decreases after jet $p_T \approx 150 \text{ GeV}$
- Data/MC agreement on 2-jet flavour fraction improved after 2 lepton cut
- Systematics in the order of few % → main syst. uncertainty: MC modelling

- b -tagging is an indispensable tool in LHC analyses
- Calibration of b -tagging algorithms necessary to correct MC eff. to data
- Performed b -tagging calibration in $e\mu + 2j$ channel of $t\bar{t}$ events
- PDF likelihood fit results with MV2c10 (fixed cut) shown
 - Good data/MC agreement observed
 - SF divergence from 1 at high jet p_T reduced by 2 lepton cut

Keep in mind: very preliminary results!



- Include 2017 data asap
 - ⇒ Need appropriate MC simulation including pileup modelling
- Studying new tagging algorithms in ATLAS
 - Deep learning tagging algorithm “DL1” (based on neural network)
 - Hybrid cut instead of fixed cut
 - (equal to fixed cut for jet $p_T < 250$ GeV; afterwards p_T -dependent efficiency)
- Implement BDT to further discriminate signal from background



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Thank you for your attention!



Backup

MC Normalisation SF

