

Performance of b-jet identification in ATLAS.

CMS Heavy flavour tagging workshop

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11 April 2018

Disclaimer on the choice of topics

All ATLAS public results regarding Flavour Tagging are available here:

<https://twiki.cern.ch/twiki/bin/view/AtlasPublic/FlavourTaggingPublicResultsCollisionData>



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I plan to discuss ...

- **a significant part of the b-tagging chain in ATLAS:**
 - flavour labeling, track association, identification algorithms, calibration.
- **emphasizing the latest developments in the group with public reference**

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I will not cover* ...

- upgrade studies
- online b-tagging performance
- c-tagging → see Andy C's talk on " $H \rightarrow cc$ in ATLAS" this morning
- more generally, new developments with not yet public reference

*sorry in advance if your favorite topic is missing, I will do my best to answer questions



Why b -jet identification so important at the LHC?

- Identifying jets corresponding to a b -quark (b -tagging) is essential to many LHC data analysis
- It is possible thanks to the high mass and long lifetime of b -hadrons

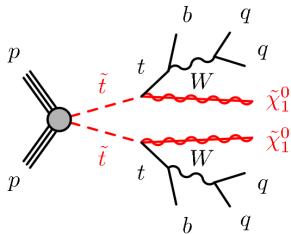
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Top Physics / New Phenomena

→ top precision measurements

→ many searches, such as searches for stop pair production (SUSY)

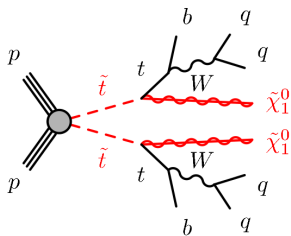


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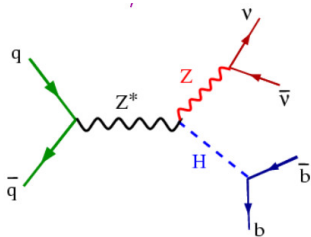
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Higgs Physics

- observation of $b\bar{b}$ decay mode
- direct measurement of the top-Higgs coupling (ttH production)

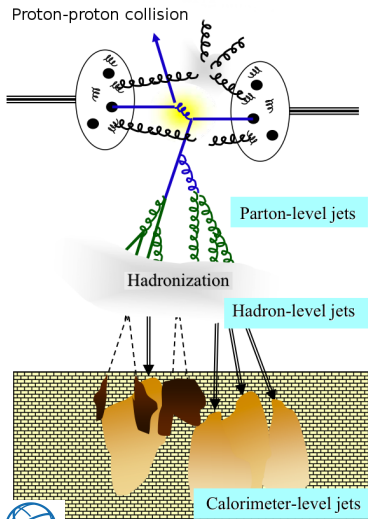


1) Flavour labeling

- How do we define a *b*-jet, a *c*-jet and a “light-flavour”-jet in our simulated samples?

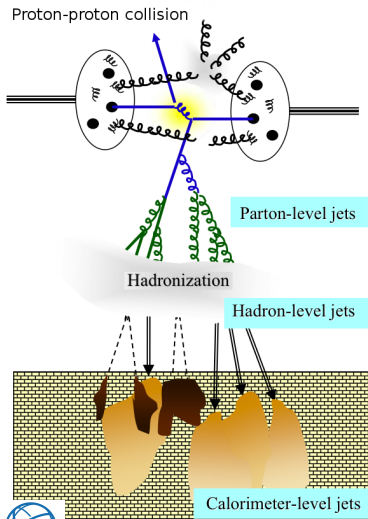


What's a hadronic jet?



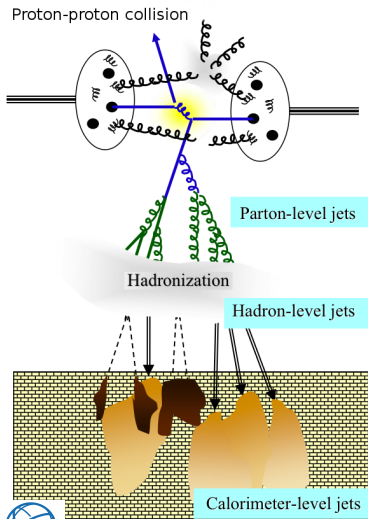
- Initial interacting partons determined from the Parton Density functions (PDFs)
- Perturbative QCD (matrix elements ME, parton shower PS), very small timescales
→ **coloured final state objects**
- Partons can be grouped together via a clustering algorithm
→ **definition of “parton-level” jets**

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- Parton → Hadrons (hadronization), other non-perturbative effects (underlying event) → **definition of “particle-level” jets**
- Experimentally, clustering based on **calorimeter energy deposits or inner detector tracks**

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$$\Delta R(\text{b-hadron, jet}) = \sqrt{\Delta\eta^2 + \Delta\phi^2} < 0.3 \quad (\text{priority to closest jet})$$

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Wrapping up, in ATLAS simulation:

b-jets are jets including at least one *b*-hadron with $p_T > 5$ GeV

c-jets do not include *b*-hadrons but include at least one *c*-hadron with $p_T > 5$ GeV

τ -jets do not include *b/c*-hadrons but include at least one τ -lepton with $p_T > 5$ GeV

LF-jets (“light-flavour”) are all the others

What about quark/gluon labeling?

A. Buckley, C. Pollard, arXiv:1507.00508 [hep-ph]

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- **QCD-aware parton labeling:** k_T algorithm on partons + prompt lepton/ γ vetoing proto-jet merging not compatible with a QCD/QED vertex
 - $k_T \sim$ “inversion of the QCD emission sequence”
 - **LO generators in good agreement**

k_T	Pythia 8	0.65	11.8	7.6	
MPI off	Herwig++	0.68	11.2	8.0	
	Sherpa	0.73	13.0	7.0	

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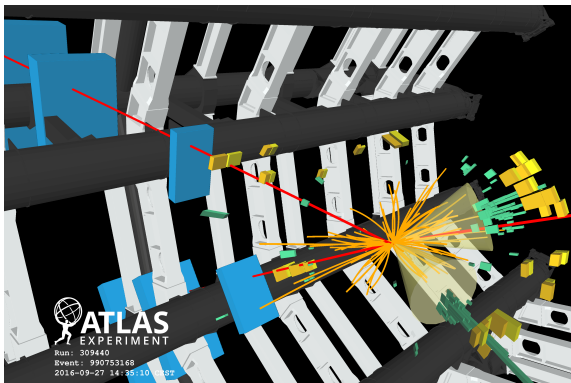
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→ $k_T \sim$ “inversion of the QCD emission sequence”
→ **computation at various perturbative order (Sherpa) in good agreement**

ME	$N_3/N_3^{2 \rightarrow 2}$	Gluon frac.	Light quark frac.	Light parton frac.	Unlabelled frac.
$2 \rightarrow 2$	1.00	62.7%	27.0%	89.6%	2.3%
$2 \rightarrow 3$	1.59	56.4%	31.4%	88.3%	2.9%
$2 \rightarrow 4$	1.79	58.3%	31.9%	90.2%	2.6%

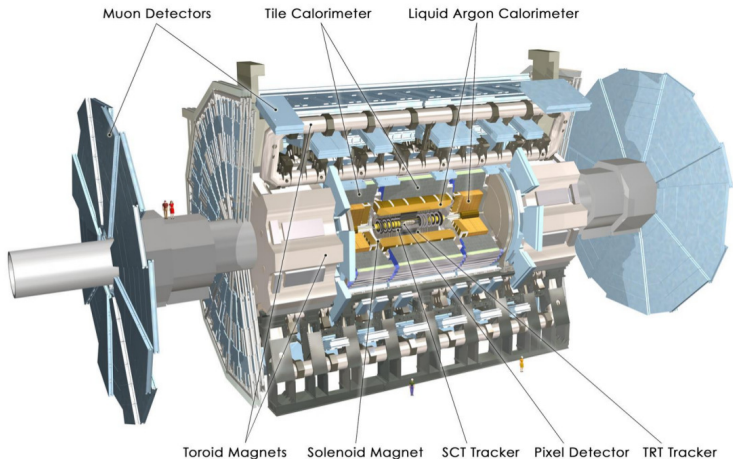
2) Jet-to-track association

- Which jet gets which tracks?

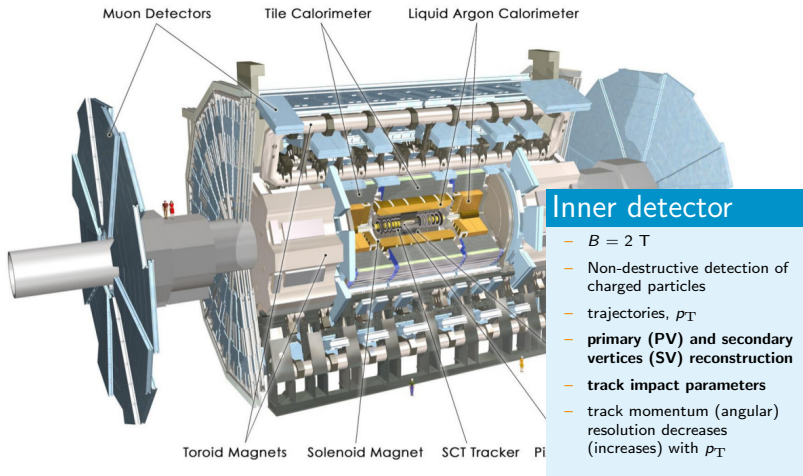


Candidate $H \rightarrow bb$ decay event with two b -jets and two muons recorded in 2016

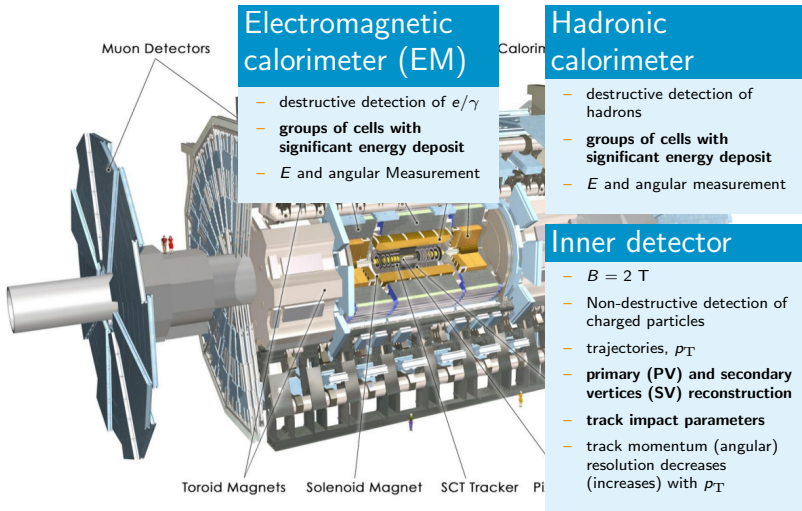
Jets and track reconstruction in ATLAS



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Muon spectrometer

- $B = 0.5$ T (mean)
- muon detection
- **not included in the ATLAS most used b-tagging algo.**

Electromagnetic calorimeter (EM)

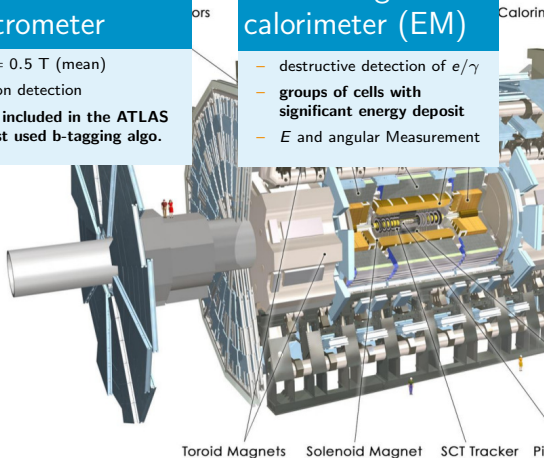
- destructive detection of e/γ
- **groups of cells with significant energy deposit**
- E and angular Measurement

Hadronic calorimeter

- destructive detection of hadrons
- **groups of cells with significant energy deposit**
- E and angular measurement

Inner detector

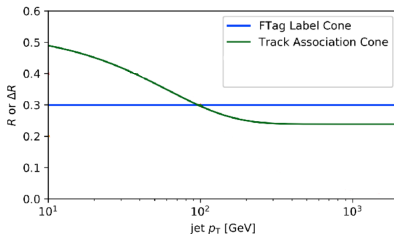
- $B = 2$ T
- Non-destructive detection of charged particles
- trajectories, p_T
- **primary (PV) and secondary vertices (SV) reconstruction**
- **track impact parameters**
- track momentum (angular) resolution decreases (increases) with p_T



Track association algorithms

See [ATL-PHYS-PUB-2017-010](#) for more details

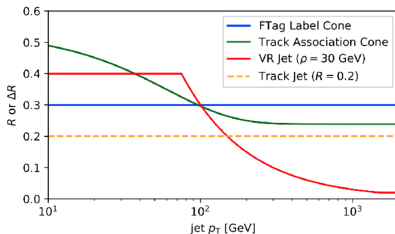
- $R = 0.4$ anti- k_t calorimeter jets are the most common jets in ATLAS
→ **shrinking cone algorithm** used to determine the tracks used for b-tagging



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- Many recent activities studying anti- k_t jets based on tracks (**track jets**)
→ **primary use case: b-tagging in dense environment** (Top, Higgs, $X \rightarrow bb$)
→ studies include e.g. $R = 0.2$ and variable- R (shrinking cone) track jets

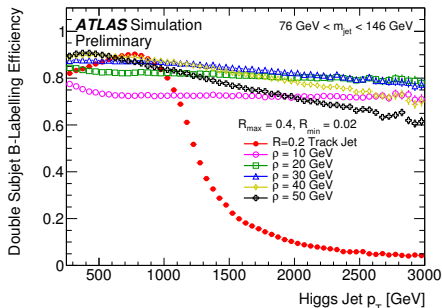
Subjet reconstruction with tracks

- Reconstruction of 1 large- R jet, **adequate constituent (subjet) reconstruction required to determine substructure accurately**
- Tracks have better angular resolution than calorimeter clusters
→ **use of track jets ghost-associated with the large- R jet**

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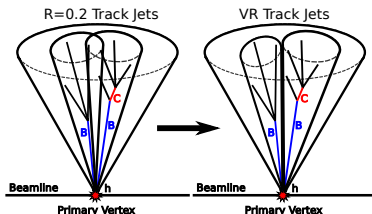
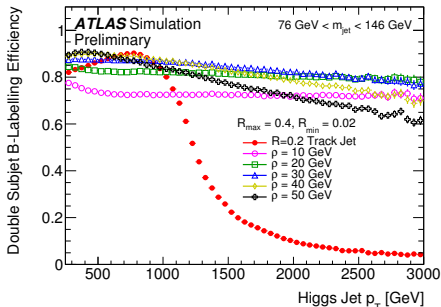
Probability to find 2 subjets truth-labeled as b



Subject reconstruction with tracks

- Reconstruction of 1 large- R jet, adequate constituent (subject) reconstruction required to determine substructure accurately
- Tracks have better angular resolution than calorimeter clusters
→ use of track jets ghost-associated with the large- R jet

Probability to find 2 subjects truth-labeled as b $R=0.2$ track jets vs Variable- R ($\sim \rho/p_T$)



- currently assessing b -tagging performance in data of such subjects
- other ideas also explored

3) b-jet identification algorithm

– Properties of b-hadrons:

- 1 **lifetime:**
 V_{cb} small, decay length $\sim 450 \mu\text{m}$
- 2 **large mass:** few GeV
- 3 **high jet momentum fraction:** $\sim 80\%$
due to b-fragmentation function
- 4 **high branching ratio to leptons:** $\sim 20\%$

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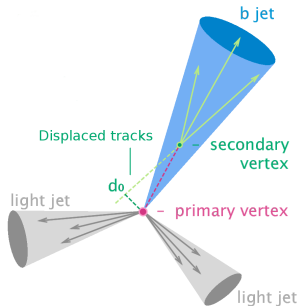
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– Experimental signatures of b-jets:

- presence of displaced tracks
- presence of secondary vertices ($B \rightarrow C \rightarrow \text{light}$)
- peculiar topology (more and higher energy tracks)
- presence of electron and muon-in-jets (not used in nominal algorithms)

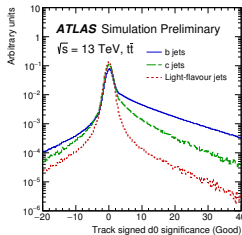
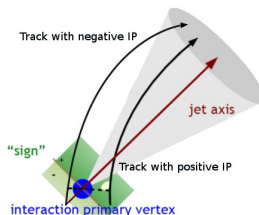


topology of *b*- and LF-jets

“Low-level” b-tagging algorithms in ATLAS

Impact parameter-based algorithm: IP2D/IP3D → used at LEP/Tevatron

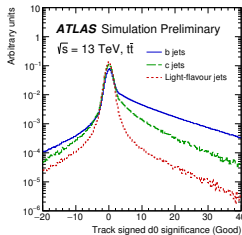
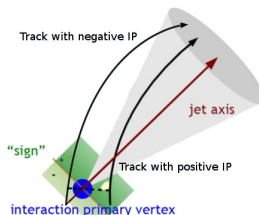
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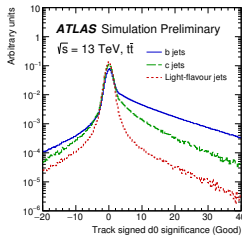
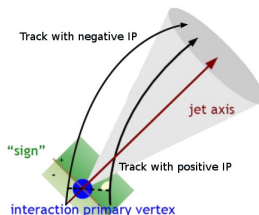
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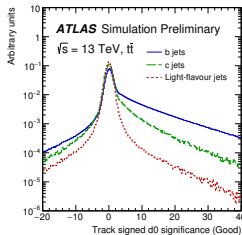
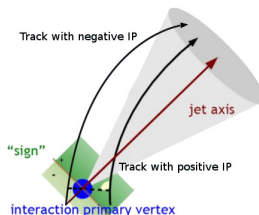
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- **large positive tails for b and c-jets**
 - 14 track categories defined
 - log-likelihoods built **per jet** from associated tracks and b -, c and LF-jet IP templates



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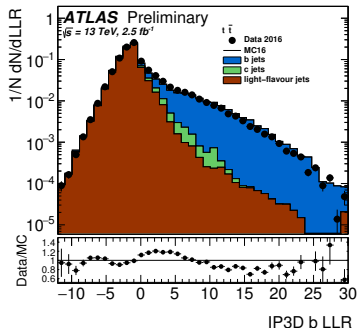
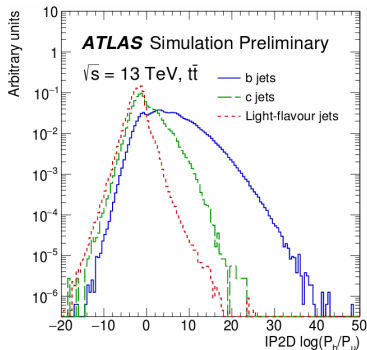
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IP2D/IP3D discriminants

see [ATL-PHYS-PUB-2016-012](#) and [ATL-PHYS-PUB-2017-013](#) for more details

$\log(P_b/P_u)$ log-likelihood discriminant for IP2D (left) and IP3D (right)



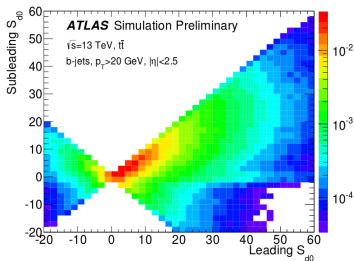
- $\log(P_c/P_u)$ and $\log(P_b/P_c)$ also defined, **total of 6 discriminants**
- **Advantage:** very inclusive, simple. **Drawback:** high sensitivity to jet axis and material interactions

New IP algorithm in ATLAS: RNNIP

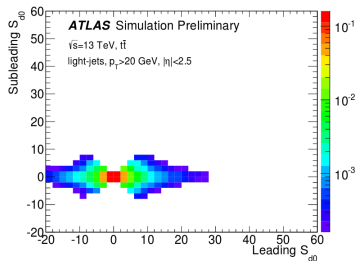
see [ATL-PHYS-PUB-2017-003](#) and [ATL-PHYS-PUB-2017-013](#) for more details

- Track impact parameters are correlated if they originate from a common decay (b -, c -hadrons), IP2D/IP3D likelihoods assume no correlation
- New IP algorithm in ATLAS learning about the correlations between tracks in b -, c - and LF-jets \rightarrow RNNIP (based on a recurrent neural network)

track IP correlations in b-jets



no correlations for LF-jets

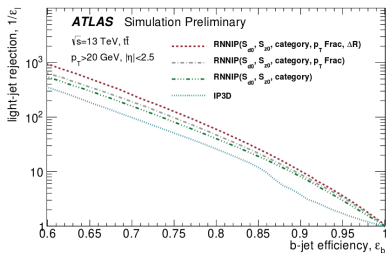


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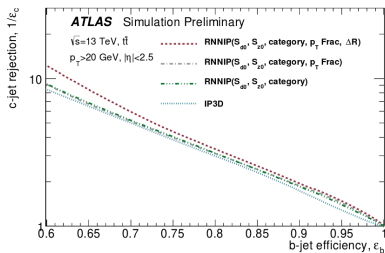
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light-jet rejection



c-jet rejection



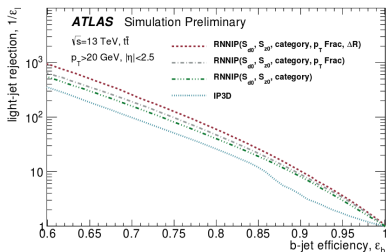
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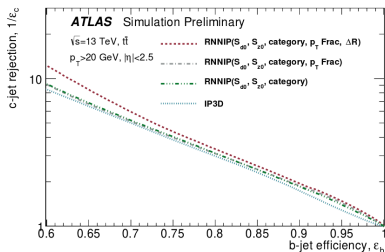
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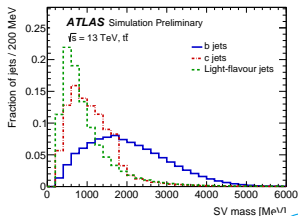
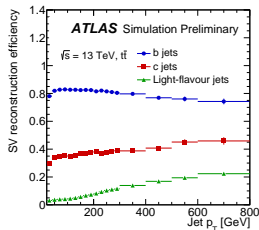


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- Performance increase if jet p_T fraction carried by track and $\Delta R(\text{track, jet})$ added
 \rightarrow studies assessing RNNIP performance in data in progress

“Low-level” b-tagging algorithms in ATLAS

Inclusive secondary vertex (SV) reconstruction: SV1

- All track pairs within a jet are tested for a 2-track vertex hypothesis
- final fit includes all tracks from 2-trk vtx
→ 1 (or 0) “inclusive” vertex per jet



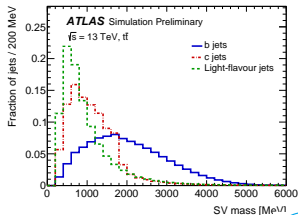
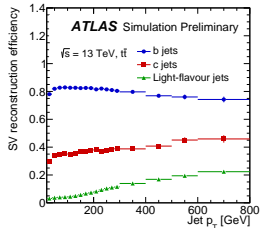
see [ATL-PHYS-PUB-2017-011](#)



“Low-level” b-tagging algorithms in ATLAS

Inclusive secondary vertex (SV) reconstruction: SV1

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→ **more SV in b- and c-jets**

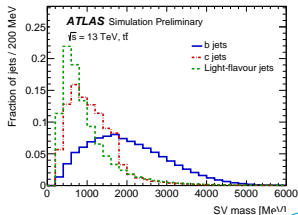
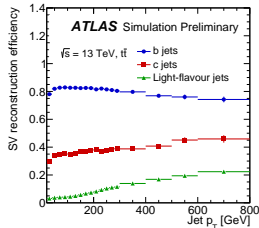


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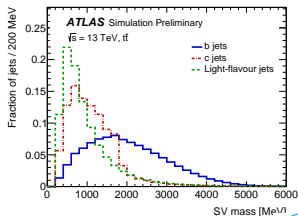
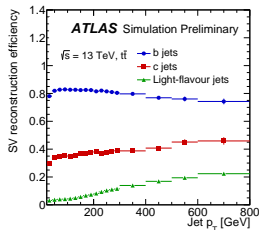


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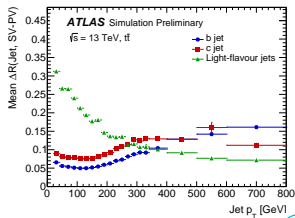
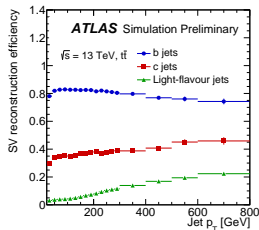


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Interesting feature: mis-alignment of the jet and PV-SV axis at high p_T for HF-jets

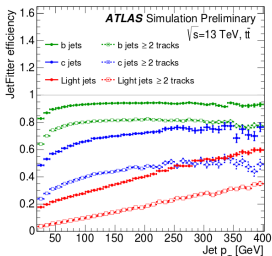


see [ATL-PHYS-PUB-2017-011](#)

“Low-level” b-tagging algorithms in ATLAS

Decay chain multi-vertex reconstruction: JetFitter

- J. Phys. Conf. Ser. 119 (2008) 03203
- exploits the topological structure of weak b - and c -hadron decays to reconstruct the full b -hadron decay chain
- b -hadron flight axis reconstructed using a Kalman filter

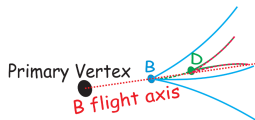


8 quantities reconstructed by JetFitter are used as discriminant

What SV1 does:



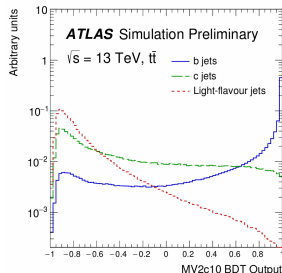
What JetFitter tries:



“High-level” b-tagging algorithm in ATLAS

$p_T^{\text{jet}} + \eta^{\text{jet}} + 3$ (IP2D/IP3D) + 8 (SV1) + 8 (JF) variables used as input to a boosted decision tree: **MV2 (multi-variate discriminant)**

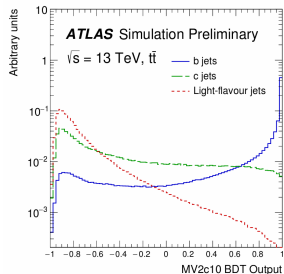
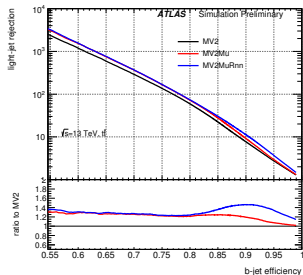
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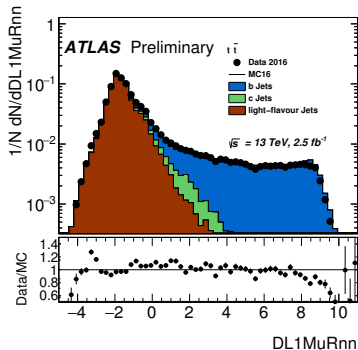
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- Performance quantified in ROC curve: **signal efficiency vs background rejection**



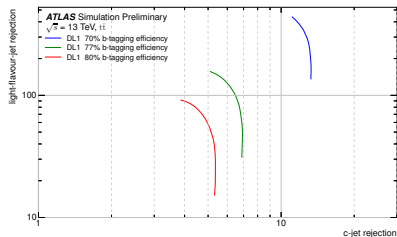
- **MV2**, main tagger used in run 2
- **MV2Mu** includes soft muon tagger output (see backup)
- **MV2MuRNN**: muons + RNNIP

New “high-level” tagger: DL1

- New high-level tagger based on a deep recurrent neural network: DL1
- fed with \sim the same information than MV2, achieve similar performance
- **but higher technology**: combined RNNIP/DL1 training, tunable c-jet fraction in the background sample without retraining, etc, possible



DL1 iso b-tagging efficiency curves

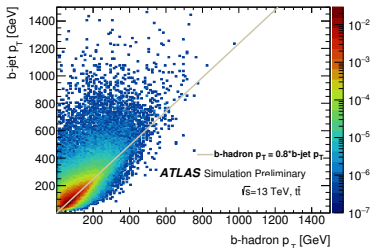


→ currently assessing DL1 *b*-tagging performance in data

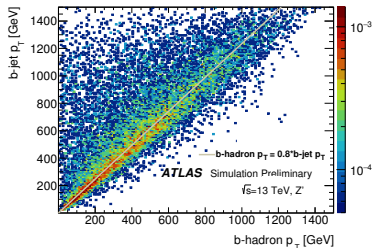
New “hybrid” training sample

- b-hadron p_T spectrum in $t\bar{t}$ intrinsically limited by $m_t \sim 175$ GeV
- for $p_T^{\text{jet}} > m_t$, jet clusters nearby hadronic activity, uncorrelated to the b-hadron (e.g. final state radiation) \rightarrow $t\bar{t}$ -based training may not be optimal
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$t\bar{t}$ simulated sample



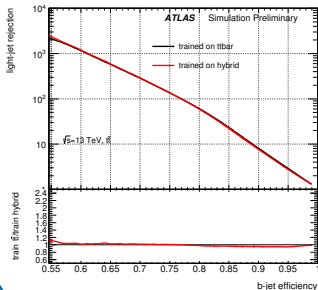
Z' simulated sample



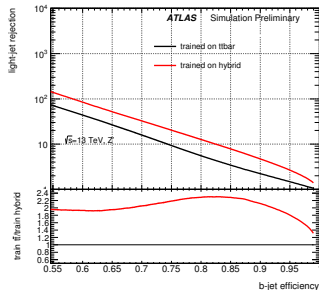
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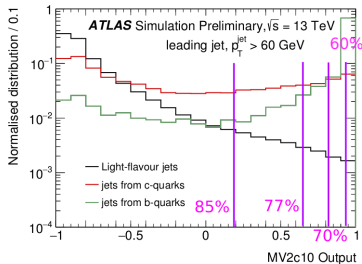


performance on Z' sample improved



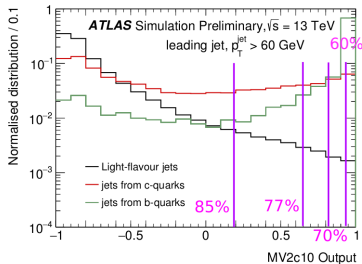
b-jet definition in collision data analysis

- **Tagger working points (WP)** defined as a certain cut on the BDT output
 - select a certain point on the ROC curves
- “fixed-cut working points”
 - constant cut value on the BDT output
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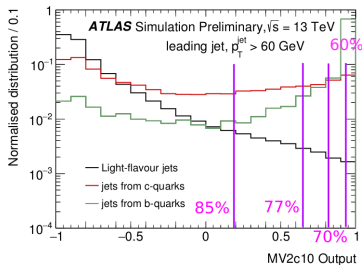
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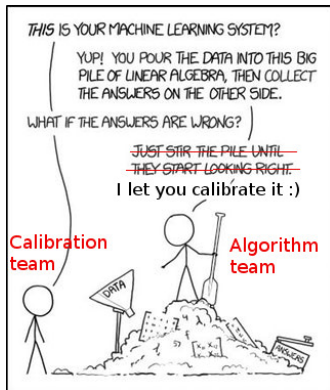
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- **Strong reasons to believe performance in simulation and data differ**
 - **for signal (i.e. true b-jets): theory modeling effects.** Uncertainty in b -fragmentation function, underlying event, ..., also pileup, tracking in dense environment at high p_T , etc.
 - **for background (i.e. non-b jets): detector effects.** Non-perfect tracker geometry, dead pixels, fake tracks from random hits, material interactions, ..., also pileup, etc.

4) b-tagging performance in data

- also known as “calibration”

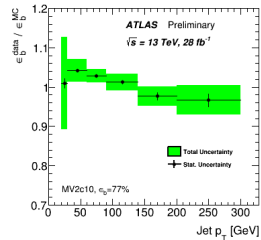
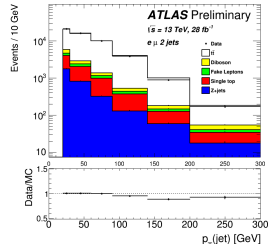


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- exactly 2 jets required to limit combinatorics to bb, bl, lb, ll

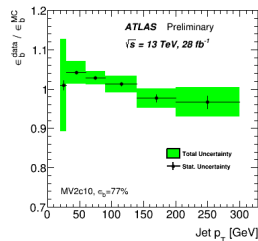
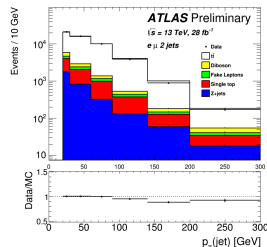


see [FTAG-2016-003](#) (paper in preparation)



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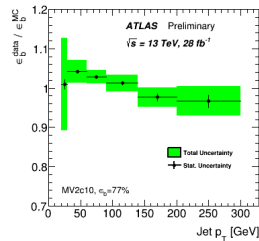
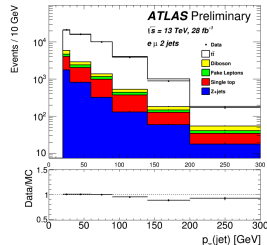
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T&P $t\bar{t}$ di- and semi-leptonic and muon-in-jet analysis also performed

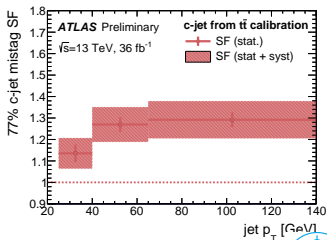
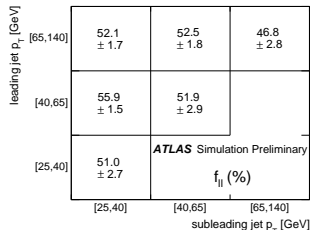


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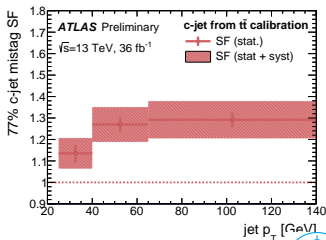
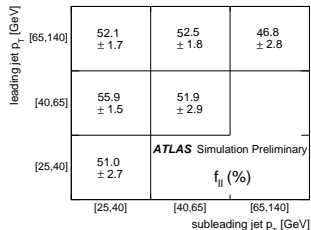
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see [ATLAS-CONF-2018-001](#)

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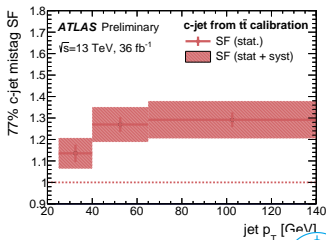
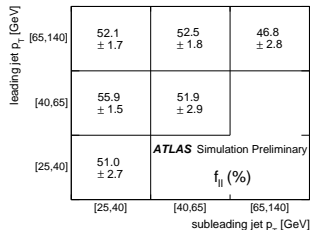


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cut & count analysis based on $W + c$ events also performed



see [ATLAS-CONF-2018-001](#)

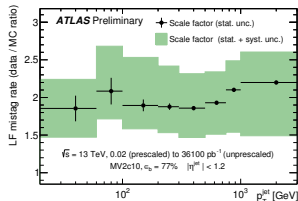
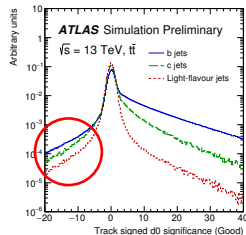


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- not achievable by regular di-jet selection:
~ 2% (5%) b-(c-)jet bef tag ... x10 after.

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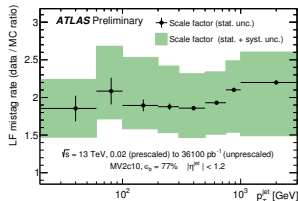
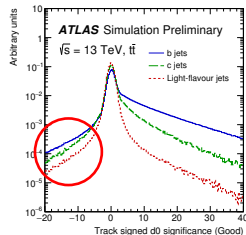
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see [FTAG-2017-002](#)
(conference note in preparation)

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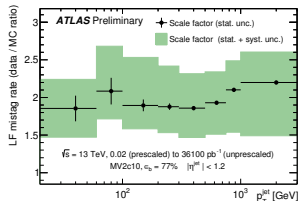
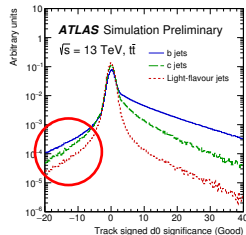
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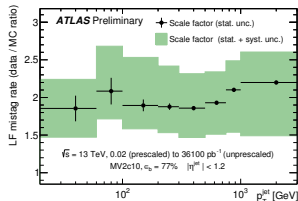
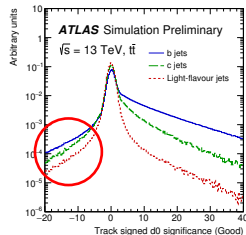


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new bottom-up approach
“adjusted-MC method” also performed



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 - b-tagging beyond the $t\bar{t}$ kinematic reach: algorithm & calibration
 - calibration of the LF-jets mistagged **not** because of track resolution effects

Performance of b-jet identification in ATLAS.

Back-up slides

Matthias Saimpert

DESY (Hamburg)

11 April 2018



QCD-aware parton labeling vertices

$$d_{ij}^{(n)} = \min(k_{T,i}^{2n}, k_{T,j}^{2n}) \Delta R_{ij}^2 / R^2 \quad D_{ij}^{(n)} = \begin{cases} d_{ij}^{(n)} & \text{if flavours QCD/QED compatible,} \\ \infty & \text{otherwise.} \end{cases}$$

k_T : $n = 1$, C/A: $n = 0$, anti- k_T : $n = -1$

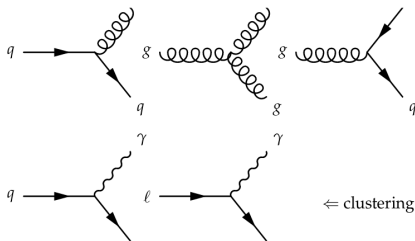


Figure 1: Feynman rule vertices used for QCD (and QED) aware jet clustering.

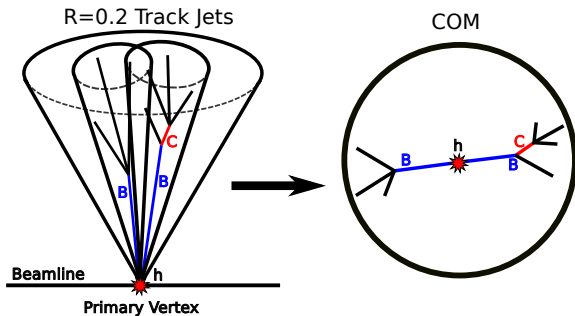
QCD-aware parton labeling with MPI on

MPI-quark particularly problematic since they can turn a gluon jet into a quark-jet

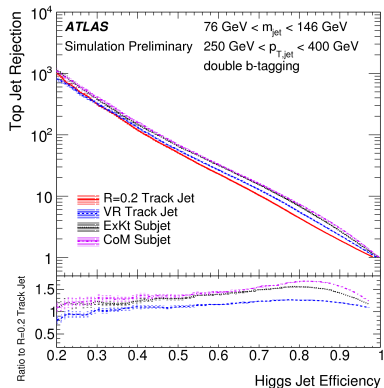
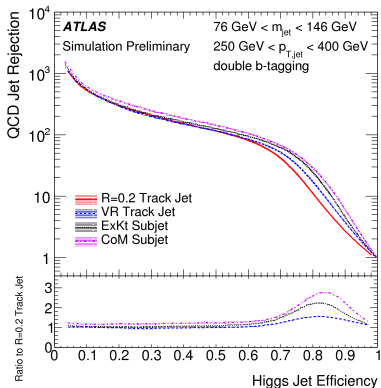
High discrepancy of Herwig++ MPI simulation with respect to the other generators

Scheme	Generator	Jets	$\gamma + \text{jet}$	
		q/g	γ/g	q/g
Max- p_T	Pythia 8	0.38	17.2	10.5
	Herwig++	0.33	7.7	4.8
	Sherpa	0.55	21.0	9.6
k_T	Pythia 8	0.80	10.4	8.2
	Herwig++	1.17	3.6	4.6
	Sherpa	0.85	10.5	7.5
anti- k_T	Pythia 8	0.79	10.2	8.3
	Herwig++	1.74	3.2	4.5
	Sherpa	0.86	10.2	7.5
Reclustered	Pythia 8	0.77	10.1	8.0
	Herwig++	1.36	3.5	4.8
	Sherpa	0.83	10.1	7.3

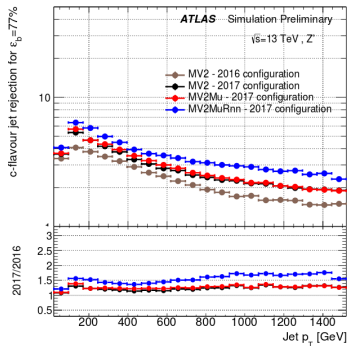
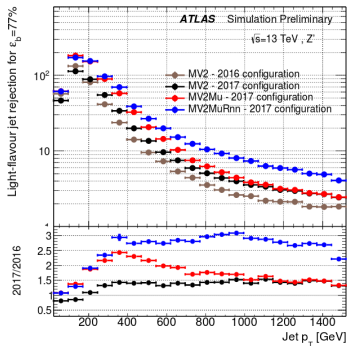
$X \rightarrow bb$ taggers: CoM sub-jets



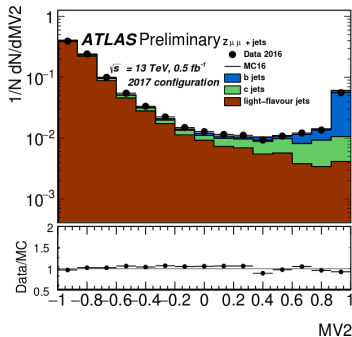
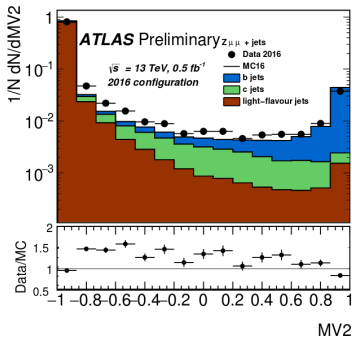
$X \rightarrow bb$ taggers: expected performance



MV2 performance in 2016 and 2017 configuration



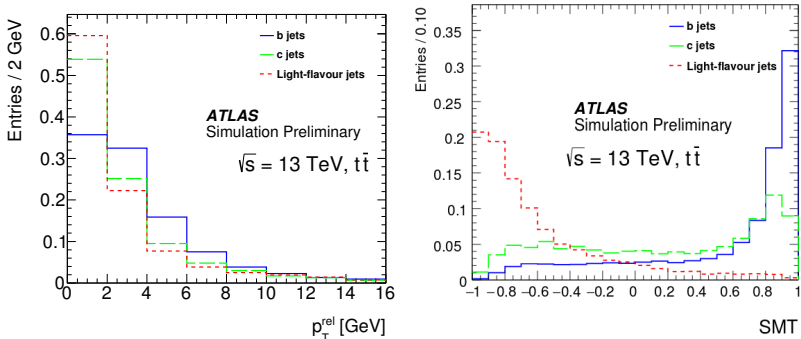
MV2 data/MC agreement with new ATLAS software (2017 configuration)



major improvements in tracking simulation in 2017 configuration
much better data/MC agreement before any calibration

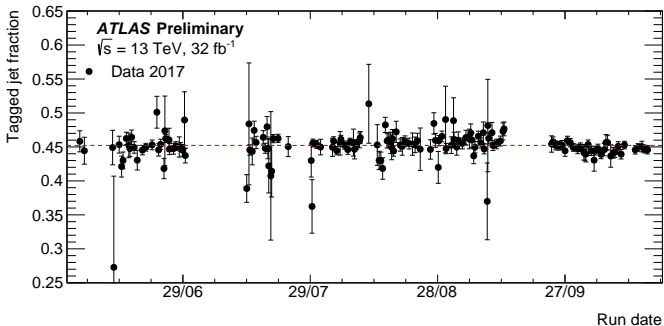
Soft Muon Tagger in ATLAS

Boosted Decision Tree discriminant based on 6 observables developed for jets including a muon



Calibration of the light flavour background challenging (track resolution effect non-dominant)

b-tagging performance stability in 2017 (1)



b-tagging performance stability in 2017 (2)

