

b-Tagging in ATLAS and CMS

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- ▶ b-tagging algorithms in ATLAS and CMS
 - Optimization of algorithms for the 2016 data taking Run
 - New developments
- ▶ Measurements of b-tagging performance in data:
 - Commissioning of b-tagging observables
 - Data-MC scale factors measurements
- ▶ b-tagging in events with boosted topologies
- ▶ Preparation for the future runs:
 - Data taking in 2017, with the new pixel Phase 1 in CMS
 - Upgrades for High-Luminosity LHC

b-Tagging in ATLAS and CMS

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Properties	ATLAS	CMS
Tracks with large impact parameters (IP)	IP2D, IP3D, JetProb TrackCounting	TCHP, TCHE, JetProb JetBProb
Secondary vertices (SV)	SV0, SV1, SV	SSVHP, SSVHE, IVF
Multiple vertices from B->C->X decay chains	JetFitter	
Soft leptons from semi-leptonic B decays	SMT, p_T Rel	Soft Lepton (SL) Taggers
Multivariate combinations	MV1c, MV2c00, MV2c10, MV2c20	CSV, CSVv2, cMVAv2, DeepCSV

- ▶ The algorithms provide a discriminant value for each jet
- ▶ Operating points defined as thresholds on the discriminant:
 - ATLAS: guaranteeing efficiencies for b-jets of 60, 70, 77 or 85%
 - CMS: reducing mis-identification rates on light jets to 0.1, 1 or 10%

RunII Algorithms in ATLAS

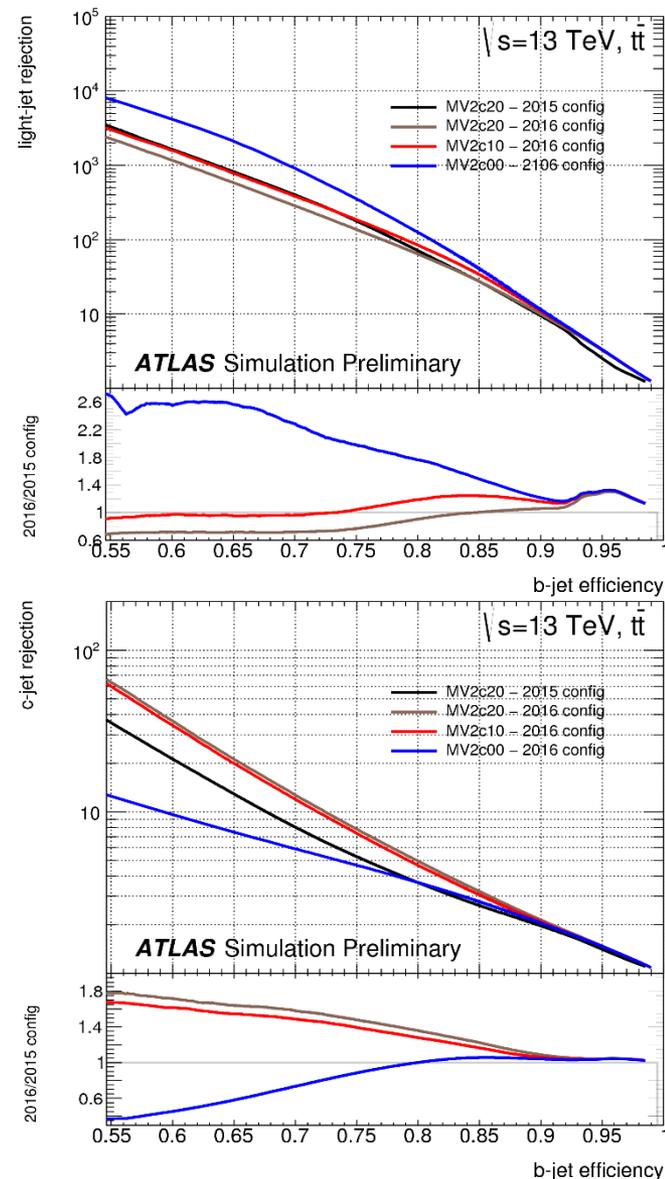
- ▶ Detector, tracking and algorithm improvements from RunI:
 - Light (c) jet rejection improved by a factor 4 (1.5-2) for a b-tag efficiency of 70%

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- ▶ Further optimization for 2016 LHC Run:

- Track selection criteria and secondary vertex cleaning in IP2D, IP3D and SV
- MVA training procedure improved
- MV2c10: similar light jet rejection, but 40% improvement in c-jet rejection for a b-tag efficiency of 70%

ATL-PHYS-PUB-2016-012

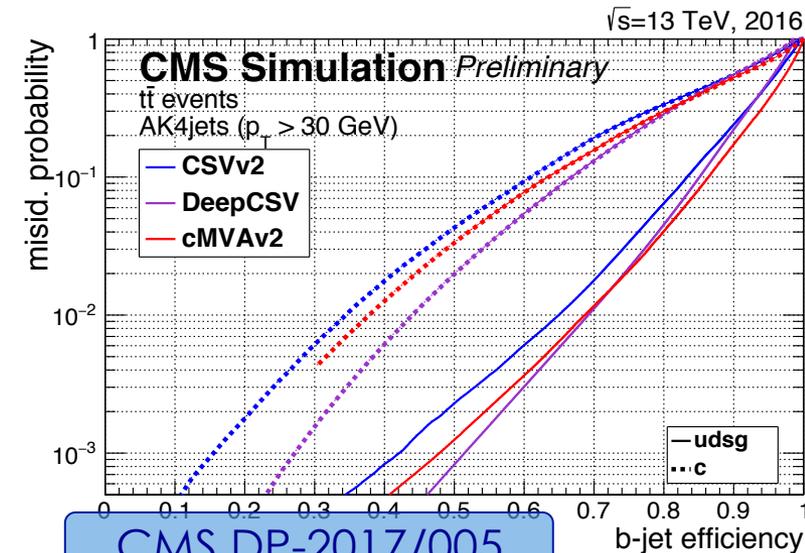
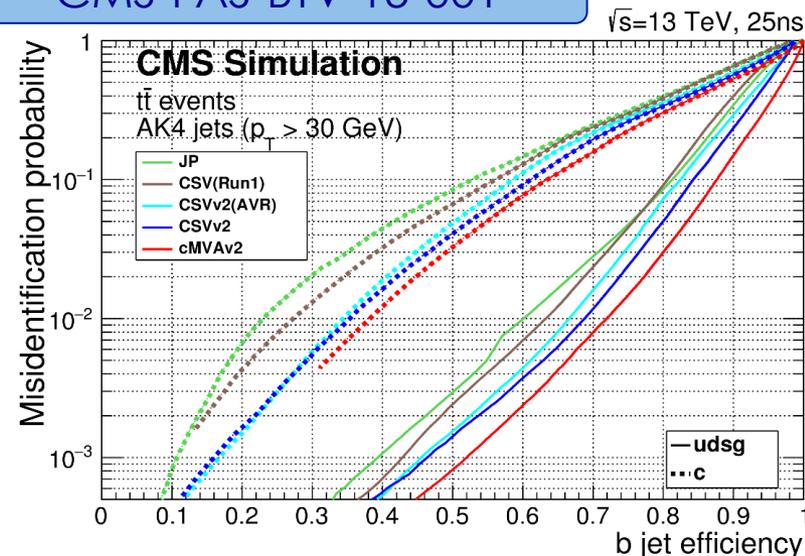


RunII Algorithms in CMS

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- ▶ The new **CSVv2** is an evolution of the RunI **CSV** algorithm:
 - Neural network instead of likelihood ratio allows to combine more variables
 - Uses Inclusive Vertex Finder
- ▶ The **cMVA**v2 tagger combines CSVv2, JP and soft lepton information
- ▶ **DeepCSV** use more sophisticated machine learning classifiers:
 - See next slide

CMS-PAS-BTV-15-001



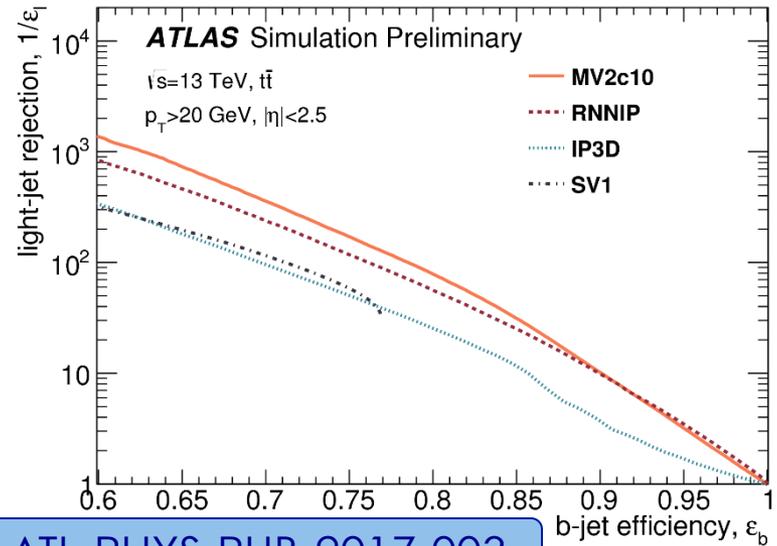
CMS DP-2017/005

- ▶ Use of more sophisticated neural network classes allows to better exploit the information available for b-tagging:
 - Can combine a large number of input features
 - Can handle more low-level information
 - Allows for multi-classification, providing an output probability for each jet flavor hypothesis

- ▶ ATLAS and CMS already developed algorithms based on:
 - Recurrent Neural Networks (ATLAS):
 - Networks with directed cycles
 - Can process arbitrary sequence of inputs
 - Deep Neural Networks (CMS):
 - Feedforward networks with many hidden layers

▶ ATLAS **RNNIP** tagger:

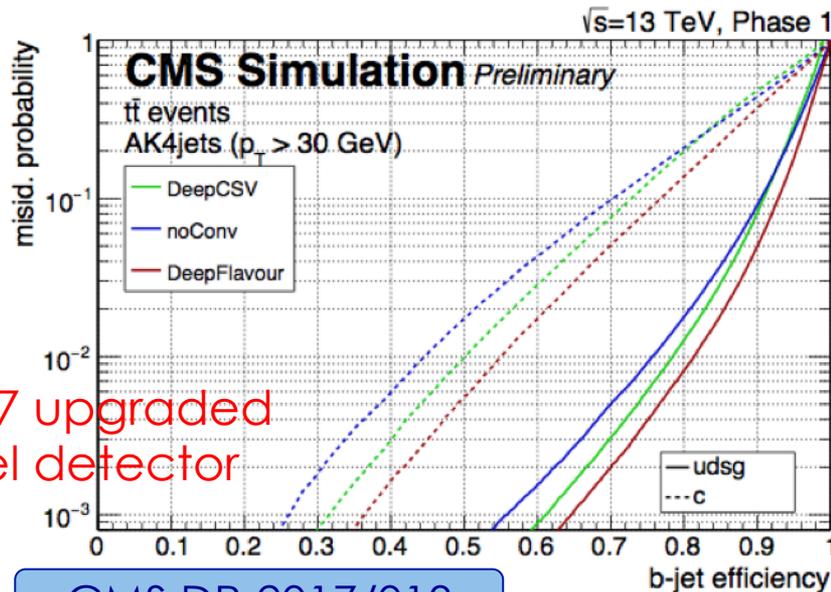
- Built using a sequence of track-by-track variables as input to a Recurrent NN
- Outperforming **IP3D**: 2.5 times its light-jet rejection for a b-tag efficiency of 70%



ATL-PHYS-PUB-2017-003

▶ CMS **DeepFlavour** tagger:

- Using properties of particle-flow candidates, and SV in the jets
- 4% absolute improvement in b-tag efficiency for a mistag rate of 0.1% against **DeepCSV**



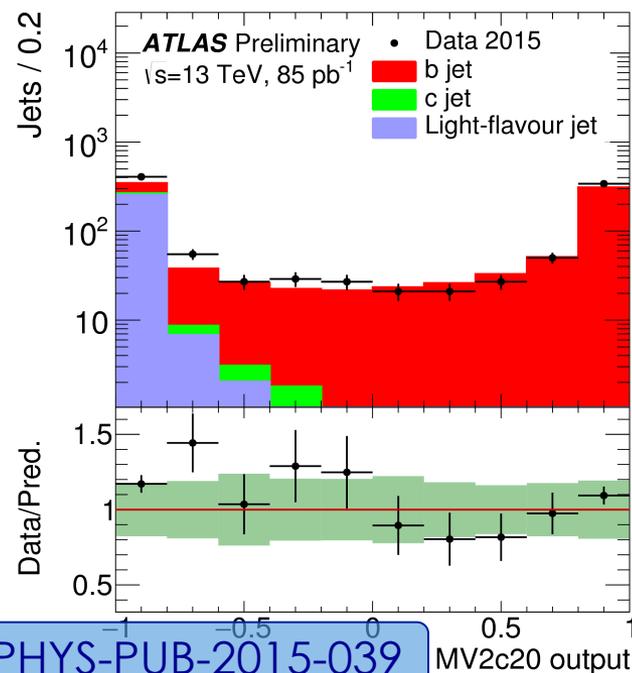
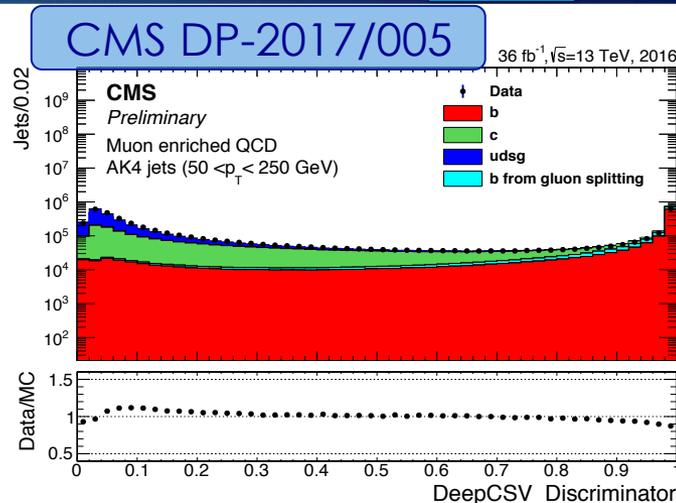
2017 upgraded pixel detector

CMS DP-2017/013

Performance In Data

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- ▶ Since simulations do not describe perfectly b-tagging observables, performance is studied in data
- ▶ Samples with different jet flavor composition are exploited to commission the algorithms:
 - Inclusive jets from QCD processes
 - Jets from QCD with an embedded soft muon
 - Top pair production events
- ▶ To correct b-tagging efficiencies in physics analysis, data-to-MC scale factors are computed for each operating point through data driven techniques

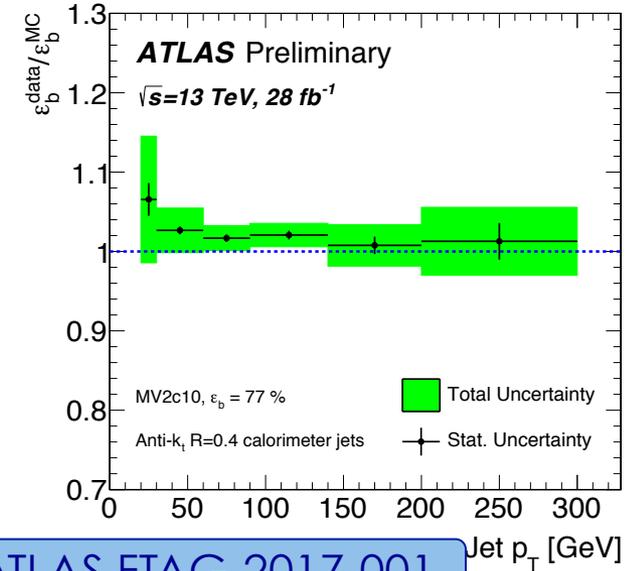
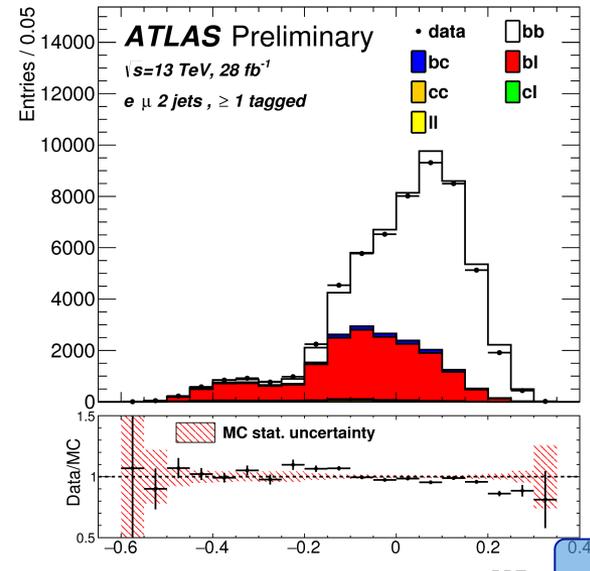
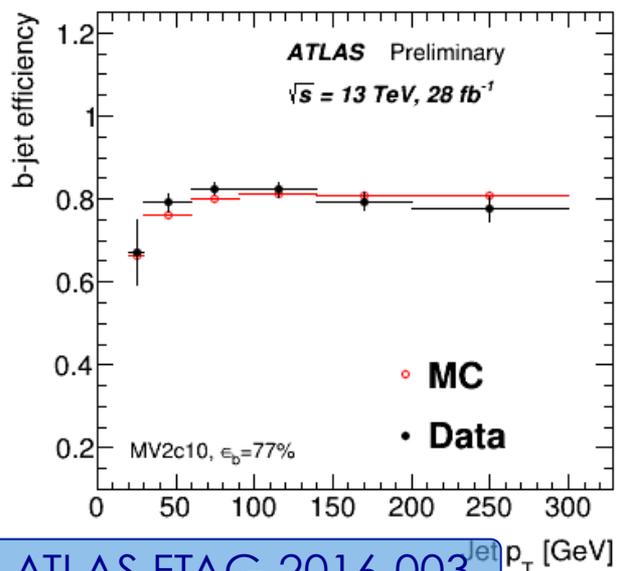


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MV2c20 output

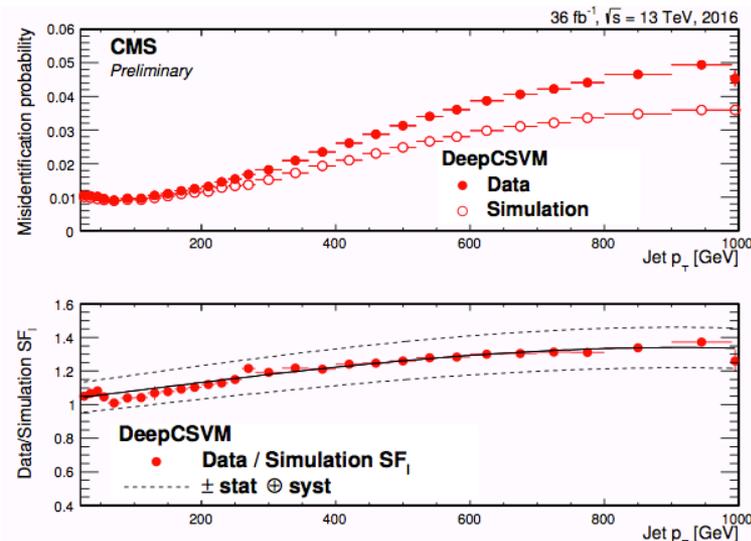
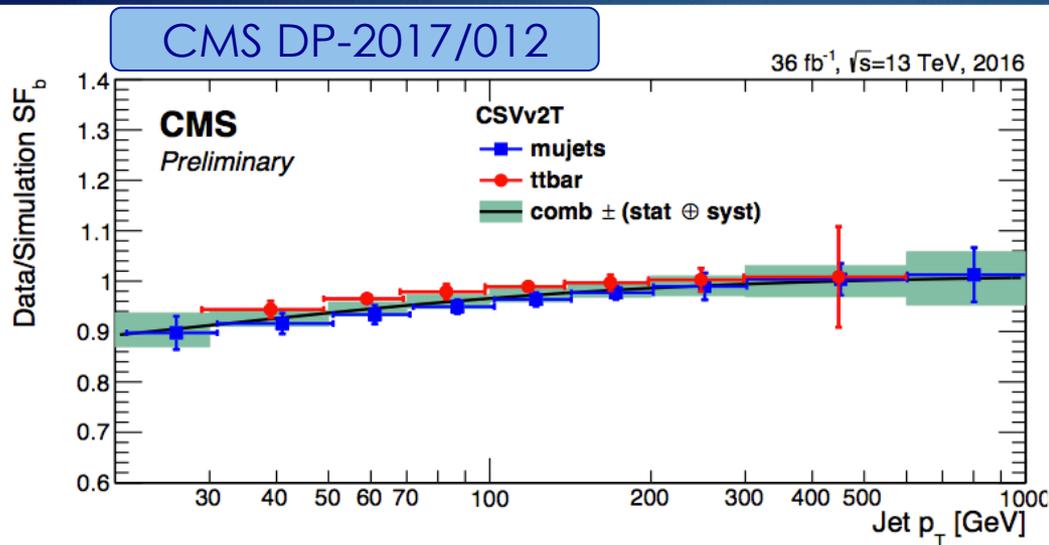
Scale Factors in ATLAS

- ▶ In RunII, scale factors measurements in ATLAS exploit the highly b-enriched jet sample from dilepton ttbar events
- ▶ Combinatorial likelihood PDF methods:
 - Efficiencies extracted from a likelihood fit to discriminant PDFs
- ▶ Tag-and-probe method:
 - Multivariate kinematic discriminator to suppress non-b jet probes



Scale Factors in CMS

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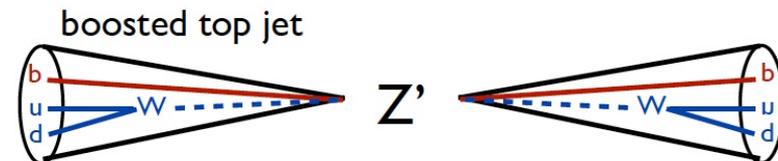
- ▶ Scale factor measurements in CMS exploits various methods:
 - QCD muon-enriched based: PtRel, System8, Lifetime Tagger
 - ttbar based: kinematic fits in dilepton and lepton+jets channels
 - Single measurements are combined to reach the best precision
- ▶ Discriminant reshaping factors are also determined in ttbar events for analyses using its whole shape in signal extraction
- ▶ SFs for light-jet mistag rate measured in inclusive jet samples

b-Tagging in Boosted Topologies

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- ▶ In high energy collisions, particles decaying to b quarks can be produced with large momentum (boosted topology):

- B decay products can overlap with particles from other jets
- Important in many BSM searches



- ▶ Both experiments developed specific b-tagging algorithms for boosted topologies, using slightly different approaches

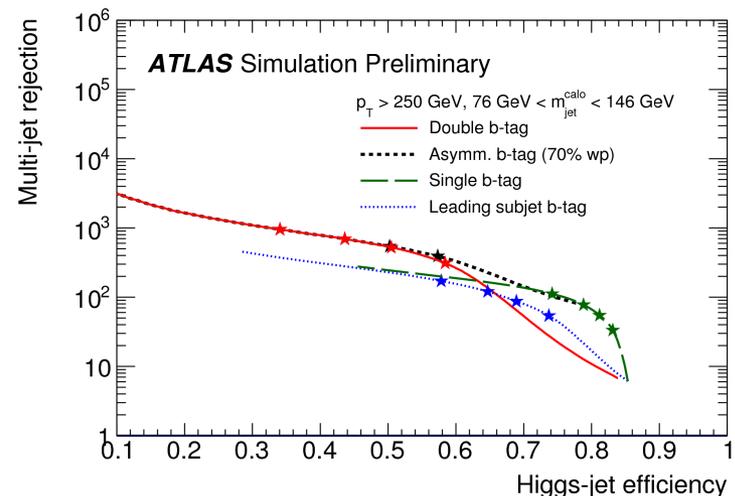
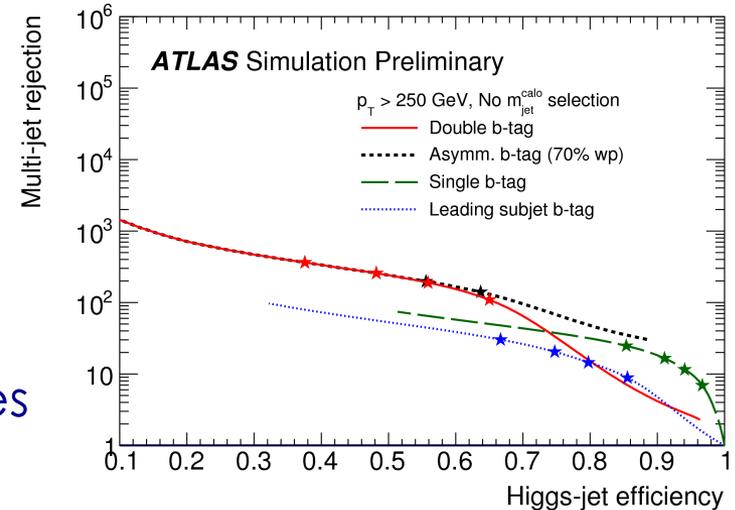
- ▶ ATLAS:

- Calorimeter large-R ($R=1.$) jets
- Trimming algorithm to discard softer components of the jet
- b-tagging applied on ghost-associated track jets ($R=0.2$)

- ▶ CMS:

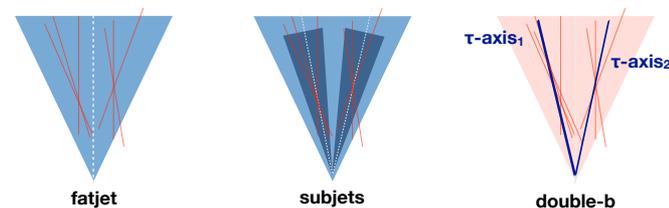
- Particle flow AK8 jets (fatjets)
- Soft drop declustering to resolve jet substructure
- b-tagging applied on tracks in fatjets and/or subjets

- ▶ Dedicated boosted Higgs boson tagger developed for RunII
- ▶ Various b-tagging options on the associated track jets compared:
 - In general, requiring two track jets satisfying (asymmetric) b-tagging requirements is more performant
 - Single track jet b-tagging becomes competitive at high efficiencies
- ▶ Additional requirements applied to enhance the performances:
 - Large-R jet mass window cut around the Higgs boson mass
 - Cut on large-R jet substructure variables (e.g $D_2^{(\beta=1)}$)



▶ Two approaches followed since RunI:

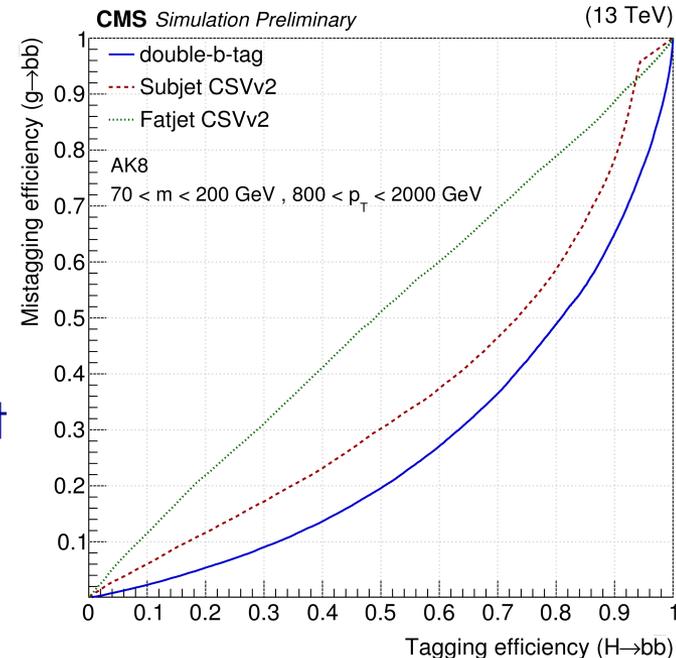
- Apply b-tagging algorithms on all the tracks in the fatjets
- Identify subjets and apply algorithms only on their tracks (still baseline for boosted top quarks)



CMS-PAS-BTV-15-002

▶ In RunII, new dedicated algorithm to tag boosted decays to b pairs:

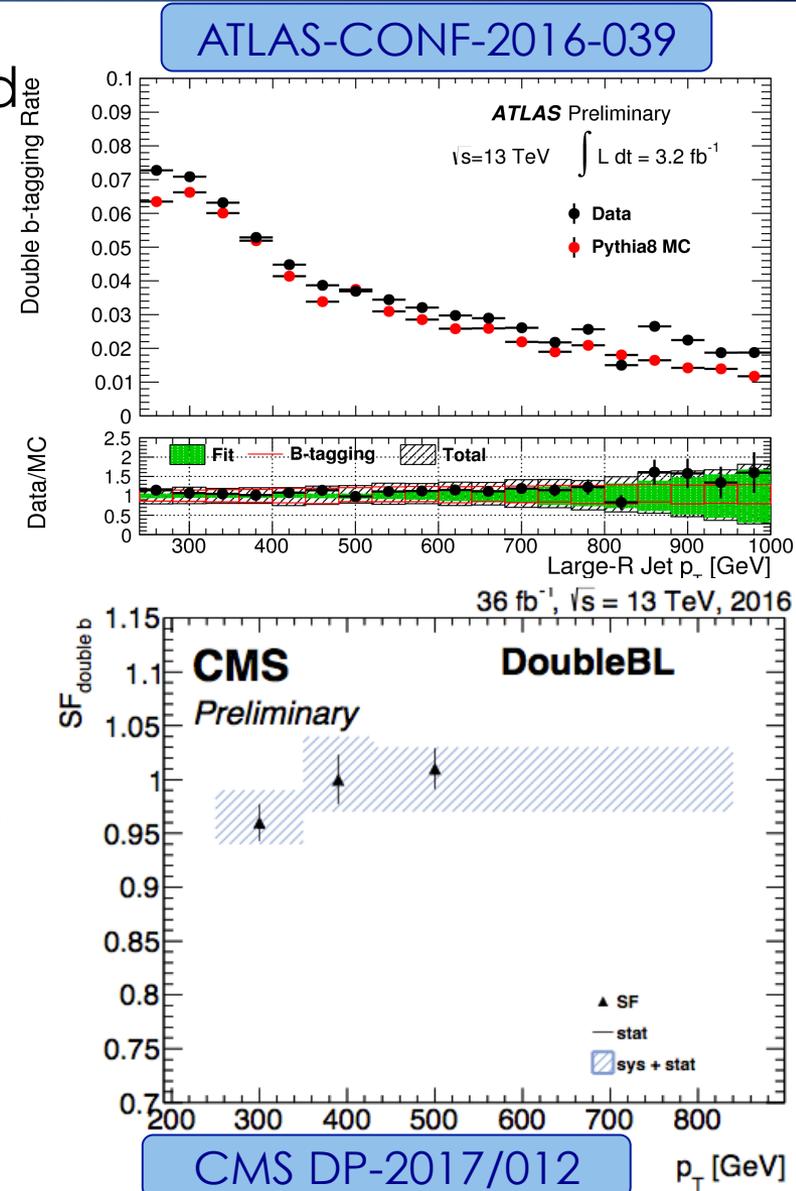
- Build observables from n-subjettiness axes to exploit the correlations between the b flight directions
- Double-b tagger outperforming fatjet and subjet b-tagging for $H \rightarrow bb$ against multijet and $g \rightarrow bb$ backgrounds



Boosted b-Tagging Validation

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- ▶ ATLAS: performance of the boosted $H \rightarrow bb$ tagger validated on candidate $g \rightarrow bb$ large-R jets:
 - Two track-jets associated, with at least one containing a soft muon
 - Observed rate of double b-tagged large-R jets agree reasonably well with simulations
- ▶ CMS: measuring specific scale factors by applying standard techniques (LT method) to AK8 jets with two soft muon-tagged subjets

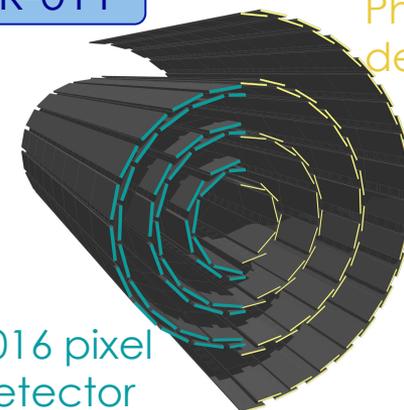


Phase 1 Upgrade in CMS

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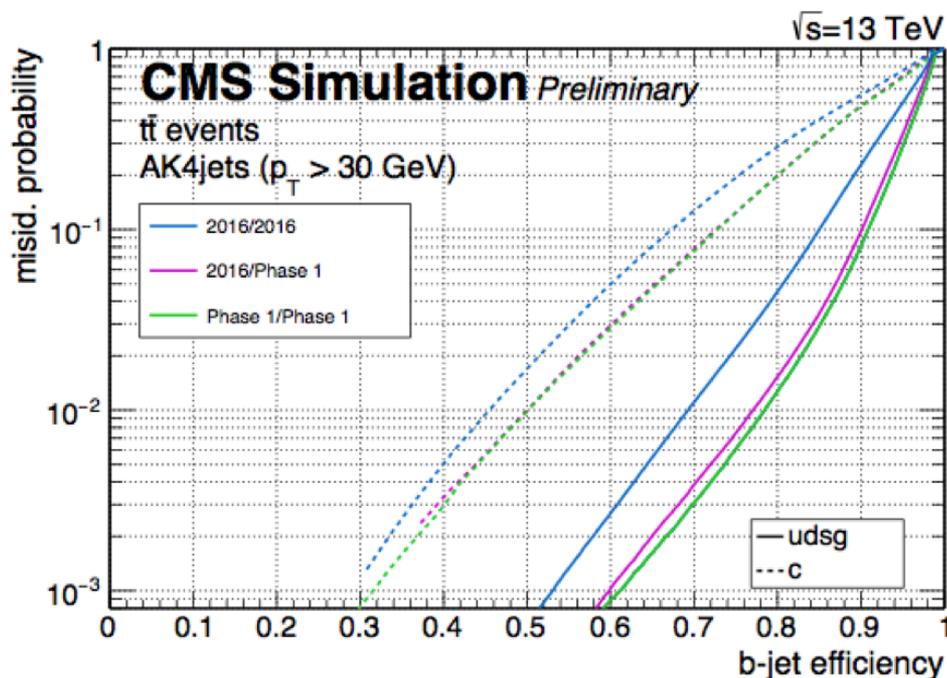
CMS-TDR-011

- ▶ The CMS Phase 1 upgrade includes a new pixel detector with an additional layer, closer to the beam spot:
 - Expected improved resolution in tracking, especially on impact parameter



Phase 1 pixel detector

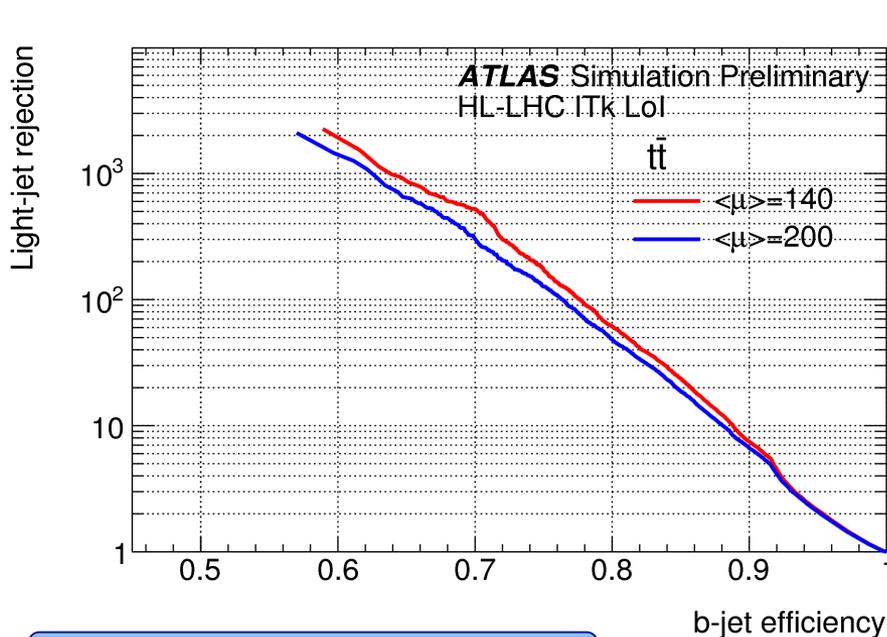
2016 pixel detector



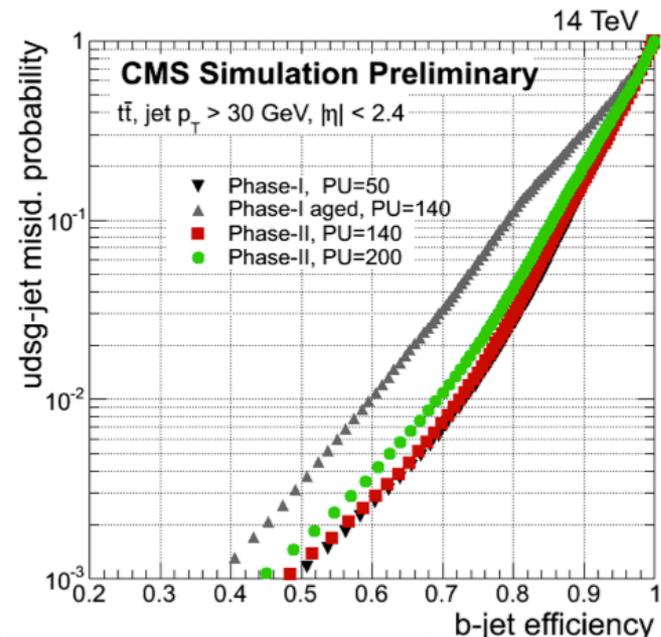
- ▶ Comparison of DeepCSV performance with 2016 detector, Phase 1 detector and 2016 training, and with Phase 1 detector and new dedicated training

CMS DP-2017/013

- ▶ Major upgrades of ATLAS and CMS detectors planned to operate during the High Luminosity (HL) LHC phase
 - Trackers will be replaced with new detector with higher granularity, radiation robustness and extended coverage
- ▶ First studies show that the b-tagging algorithms can operate in the complex high PU environment expected during HL-LHC



ATL-PHYS-PUB-2016-026



CMS DP-2016/065

Conclusions

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- ▶ b-Tagging is a fundamental tool in most physics analyses
- ▶ Both ATLAS and CMS reached a significant improvement on their algorithms in RunII, and new promising ideas for further developments are being explored
- ▶ Not only algorithms, but also the measurements of their performance on data had benefited from new ideas (and of increased sample statistics) in 2016
- ▶ Techniques are being extended to cover more specific topologies becoming ever more important with the increase of the LHC collisions center-of-mass energy
- ▶ More challenge ahead: already working to maintain b-tagging a successful tool in the next decade of data taking

Backup Material

Identification of c-Jets

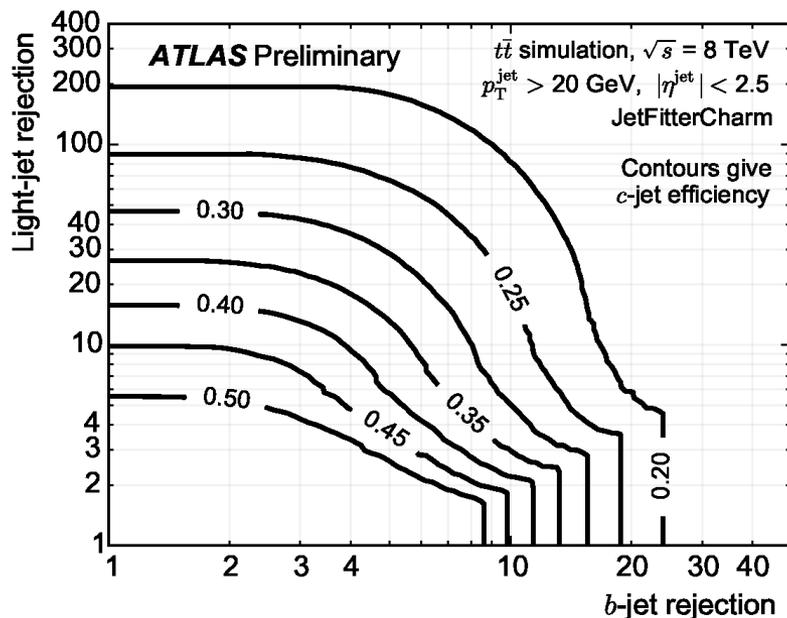
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- ▶ ATLAS Run1 JetFitterCharm uses a design similar to MV1:

- NN combination of a modified version of the IP3D, SV1 and JetFitter taggers

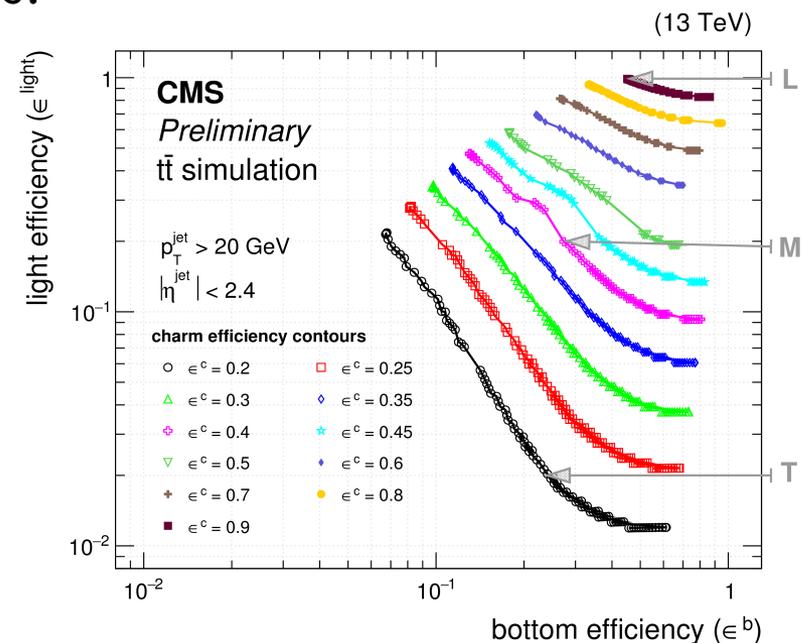
- ▶ Operating points defined by 2D cuts on c-jets vs light-jets and c-jets vs b-jets discriminants:



CMS-PAS-BTV-16-001

- ▶ The CMS c-jet tagging algorithm uses a design similar to CSVv2:

- Adding soft lepton information to add more observables and more jet categories

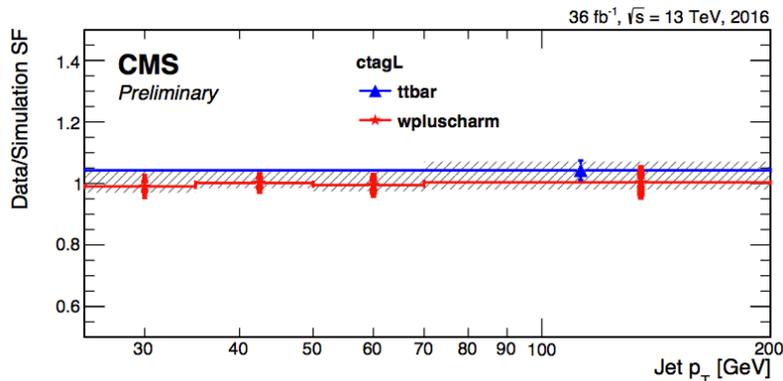
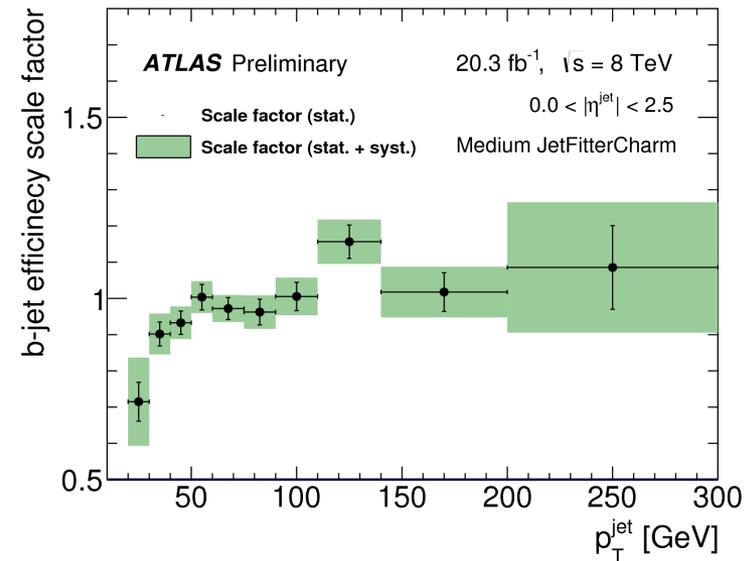


Scale Factors for c-Taggers

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- ▶ In ATLAS, efficiencies for c-jet identification are measured in a multijet event samples where jets contain D^* mesons

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CMS DP-2017/012

- ▶ CMS: performance measured in two c-jet enriched samples:
 - ▶ Wc events, selected by requiring a soft muon in the c-jet
 - ▶ ttbar events in lepton+jets final states

c-Tagging with Multiclassifiers

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- ▶ DeepCSV is already outperforming the dedicated c-tagger

