b-Jet tagging in proton-(anti)proton collisions

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Contents:

- tagging algorithms
- tagging performance calibration

Introduction



Physics goals

The largest branching fraction for a M_H =126 GeV Standard Model Higgs boson is H \rightarrow bb

- exploit in the search
- information on coupling to fermions

Identifying b-jets is fundamental to measuring $|V_{tb}|$

 notably EW production of single top quarks (in association with b quarks)

Many SUSY searches focus on the third quark family being lightest

• exploit in $\tilde{g} \rightarrow \tilde{b}b, \tilde{b} \rightarrow b\chi^0$



Tracking detectors

Essential: superb and redundant tracking!

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- trend continued by ATLAS, CMS: yet more silicon strips; pixels!
- essential to reduce sensitivity to pile-up



Tagging preliminaries

Primary vertex selection: typically based on tracks' p_T

- ATLAS, CMS: select PV with highest Σp_T^2
- will need to revise in view of further increased pile-up starting from 2015
- $Z \rightarrow \mu\mu$ candidate with 25 reconstructed primary vertices



Track selection:

- generally use a fixed cone around calorimeter jet direction
- but e.g. ATLAS exploits collimation at high jet pT: smaller cone size
- rejection of tracks from long-lived light hadrons
 & material interactions
- attempt to identify these explicitly



 $\boldsymbol{\gamma}$ conversions in the D0 tracker

Impact parameter tagging

Time-honoured: impact parameter tagging

- based on (usually physics-signed) impact parameter significance $S_{d0} \equiv d_0/\sigma(d_0)$
- 2D or 3D
- Simplest forms: track counting (e.g. number of tracks with significance above a specified minimum)
- nice CMS variant: use 2nd or 3rd highest significance
- less prone to single outlier tracks
- JetProb: compute joint compatibility of jet's tracks with the PV
- use of resolution function derived from d₀ < 0 tracks: (partly) self-calibrating!

ging	Displaced
c	Tracks
Jet	Vertex
Primany	do
Vertex	
	let

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data

b guark

c quark

20

b from gluon splitting

uds quark or gluon

25

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$$\mathcal{P}_{\text{trk},i} = \int_{-\infty}^{-|\mathcal{S}_{d_0}^i|} \mathcal{R}(x) dx,$$

$$\mathcal{P}_{\text{jet}} = \mathcal{P}_0 \sum_{j=0}^{N-1} \frac{(-\ln \mathcal{P}_0)^j}{j!}, \quad \mathcal{P}_0 = \prod_{i=1}^N \mathcal{P}_{\text{trk},i}$$





Secondary vertex tagging

Time-honoured: allow for significantly higher purity than impact parameter tagging

- typically use"build-up" algorithms: start with 2-track vertex candidates, attempt to add other tracks
- cleaning stages to remove tracks originating from the primary vertex, fake tracks, ...
- access to interesting kinematic information!
- e.g. charm; $g \rightarrow bb$

1800

1600

1400

1200

1000

800

600 400

200

Data/MC





A more unified approach

The assumption of a single secondary vertex is not always valid!

- b \rightarrow c \rightarrow I decays: also charmed hadrons live long!
- JetFitter approach used in ATLAS (similar to earlier SLD algorithm): assume vertices are aligned
- allows for significant efficiency increase

After reconstruction:

- categorise by track and vertex multiplicity
- construct likelihood ratio from discriminating variables
 \$\mathcal{P}_b\$

 $\overline{\mathcal{P}_b + \mathcal{P}_c + \mathcal{P}_{\text{light}}}$

Similar features used in CMS SV reconstruction



Primary Vertex B B flight axis

Multivariate tagging

Recent example: CDF's Higgs-Optimized b-Identification Tagger (HOBIT)

- combine 25 variables sensitive to b \leftrightarrow non-b differences in a neural network
- secondary vertex related, impact parameter related, soft muons
- including outputs of earlier neural networks
- significant improvements, can be optimised for specific physics use cases



 versatility exploited in ATLAS by optimisation for charm (instead of light-jet) rejection

Performance estimates

Impressive range of algorithms, but MC estimates should be checked!



Calibrations of efficiency for b / c / light-flavour jets to satisfy given cuts on tagging outputs

result typically reported as data/MC efficiency scale factors

Muon-based calibration (1)

Most focus in b-jet tagging is on lifetime tagging

 semileptonic decays (especially involving muons) suffer from low branching fractions; but useful in calibrations (see later) and for specific physics analyses

But jets with associated muons are very useful for calibration purposes (in addition to providing specific analysis use cases)

- most commonly used: p_{T,rel} (good at low p_T)
- also muon impact parameter (better at high p_T)



Muon-based calibration (2)

Reduced dependence on MC: System8 (pioneered by D0)

- use 3 (essentially uncorrelated) cuts each discriminating between b-jets and c/l-jets: the tagger under consideration (LT), a p_{T,rel} cut (MT), and an away-jet tag (n → p): 2³ = 8 combinations is solve for 8 unknowns
- need to lump together charm and light-flavour jets
- correlations (mostly inferred from MC) are typically small.
 Exception: charm purification by away-jet tag (difficult to estimate but extracted b-jet efficiency not very sensitive)

$$\begin{aligned} \alpha_1 &= \varepsilon_b^{LT,MT,n} / (\varepsilon_b^{LT,n} \varepsilon_b^{MT,n}) \\ \alpha_5 &= \varepsilon_b^{MT,p} / \varepsilon_b^{MT,n} \\ \alpha_6 &= \varepsilon_b^{LT,p} / \varepsilon_b^{LT,n} \\ \alpha_7 &= \varepsilon_b^{LT,MT,p} / (\varepsilon_b^{LT,p} \varepsilon_b^{MT,p}) \end{aligned}$$



Secondary Vertex

do

Jet

Primary Vertex Displaced Tracks

Both results need to be extrapolated to inclusive b jets

assume same data/MC scale factors hold

Reference lifetime tagger

Basic problem with muon-based calibration: $p_{T,rel}$ spectra for b-jets and lighter flavours become similar at high $p_T \implies$ not useful for $p_T > 200 \text{ GeV}$ CMS are using an alternative method: JetProb template fitting

- templates still partly based on MC but under improved control:
- self-calibrating feature for light-flavour jets
- JetProb calibration



tt-based calibration

For the LHC experiments only!

- greatly increased tt production cross section (from ~ 7 pb to ~ 200 pb)
- relatively clean source of b jets (assuming $B(t \rightarrow Wb)=I$)

Exploited using a variety of methods, both in dilepton & lepton+jets events

dilepton sample:

- either count fraction of tagged jets
- or fit to tagged jet multiplicity distribution (predictions depend on ε_b)



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lepton+jets sample:

- enhance tt content using kinematic selection, count fraction of tagged jets
- or apply kinematic fit, subtract wrong combinations & count



No extrapolation to inclusive b jets; but modelling uncertainties

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ATLAS: use $D^{*+} \rightarrow D^{0}(\rightarrow K^{-}\pi^{+})\pi^{+}$ decays to remove light-flavour jets

- vertexing but without track impact parameter cuts / fit χ^2 cuts to minimise biases on tagging measurements
- minimum $E(D^*)/E(jet)$: suppress combinatorics + b \rightarrow D^{*}X decays
- estimate remaining b-jet component using pseudo-cτ fit (before tagging)
- subtract (small) remaining) b-jet contribution using (corrected) simulation



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Mis-tag rate calibration

Negative-tag calibration: exploit symmetry of track resolution functions

... and the availability of d < 0 tracks, L < 0 secondary vertices

Determine negative-tag rate in data, correct for sources of asymmetry (MC)

- heavy-flavour contamination of negative-tag data
- long-lived hadrons in light-flavour jets

Substantial MC dependence

- but can use e.g. measured heavy-flavour fractions, strange particle production rates; quality of detector simulation
- CMS example: good agreement over a large p_T range

 $\epsilon = \epsilon^- \cdot R_{\rm hf}^- \cdot R_{\rm ll}$

Summary & Outlook

Recent years have seen significant advances in b-tagging:

- Tevatron experiments have continued the improvements of their b-tagging algorithms
- The LHC experiments benefit greatly from detector advancements, very good detector simulations, and much higher (tt) cross sections

As a result, powerful and robust tagging algorithms exist, and their performances are characterised by a wide variety of calibration methods

The (LHC) work isn't done

 the next challenge will be to accommodate the further luminosity increase after the upcoming shutdown

