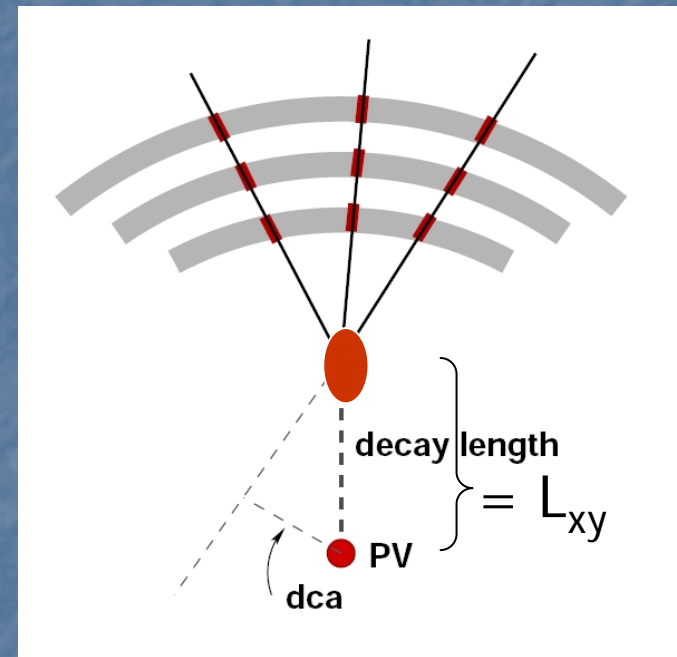


# Calibration of b-tagging at Tevatron

1. A Secondary Vertex Tagger
2. Primary and secondary vertex reconstruction
3. Tagger characteristics
4. Determination of "b-tagging efficiency"
5. ("c-tagging efficiency")
6. Determination of "mistag rate"
7. Systematics

# Secondary Vertex Tagger algorithm

- Explicitly reconstruct secondary vertices (other options are counting displaced tracks,...)
- Used for most  $D\bar{0}$  top results  
 $B(t \rightarrow Wb) \sim 1$
- Similar algorithm used by CDF
- Requires the position of the primary interaction (primary vertex or PV)



# The Secondary Vertex Tagger Algorithm (SVT)

## Three steps

- I. Reconstruction and identification of a primary vertex (PV)
- II. Reconstruction of track based jets ("**track-jets**")
- III. Secondary vertex finding

## Step I: determine PV on a per-event basis

1. Fit all well reconstructed tracks to a common point of origin,
2. Remove tracks with too high  $\chi^2$  contributions,
3. Repeat with remaining tracks,
4. **Select main PV with  $p_T$  distribution least consistent with min. bias (DØ),**
5. **Select main PV closest to high  $p_T$  lepton or PV highest scalar sum of track  $p_T$  (CDF).**

## Step II: track based jets or "track-jets"

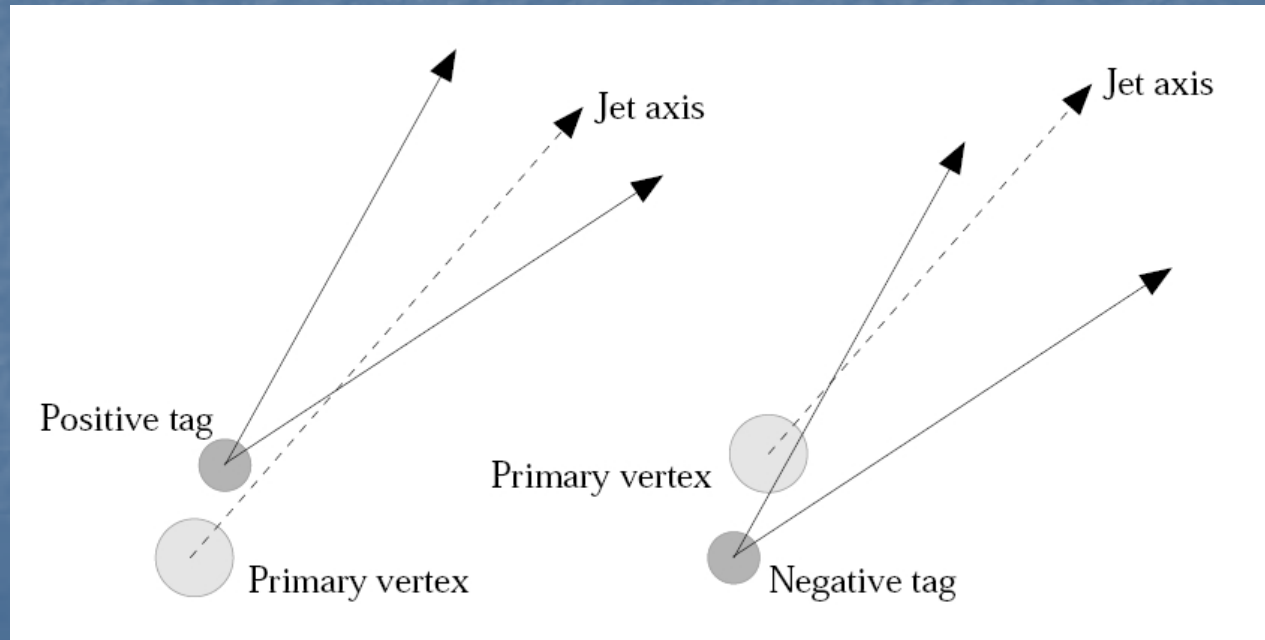
1. Pre-clustering: make precluster in  $z$  (along beam axis) of tracks that are nearby in  $z$ . Start from highest  $p_T$  tracks.
2. Track selection: associate each precluster to the closest PV, use tracks that have  $p_T > 0.5$  GeV,  $\geq 1$  hit in the most precise section of the silicon, small  $dca$  and  $z_{dca}$
3. From the preclusters, the tracks are clustered with a simple cone algorithm, with a track seed of  $p_T > 1$  GeV.

Track-jets useful in many other situations...

## Step III: Secondary vertex finding

1. Start from seed vertices in each track-jet (i.e. all pairs of tracks)
2. Add tracks to seed vertices if there  $\chi^2$  contribution is not too large
3. Select vertices with  $\geq 2$  tracks,  $|L_{xy}| < 2.6$  cm (within first silicon layer!),  $L_{xy} > n \times \sigma(L_{xy})$ , (adjust  $n$  to required rejection),  $\chi^2, \dots$  (2 steps in CDF.)

"b-tagged" = there is  $\geq 1$  SV within  $\Delta R=0.5$  of the calorimeter jet.



# Tagger characteristics

- Probability to tag a b-jet =  
"b-tagging efficiency"
- Probability to tag a light jet (g, u, d, s)  
"mistag rate"
- Probability to tag a c-jet  
"c-tagging efficiency"

These parameters are in general functions of the jet  $p_T$  and  $\eta$ ,  
Could also be dependent on the PV position, the luminosity, run  
range ...

# Probability to tag a b-jet

To decouple from detector issues, define (CDF and DØ)

- I. Taggable jets, (experiment wide definition)
- II. Tagged jets

A calorimeter jet is taggable if:

1.  $E_T > 15$  GeV,  $|\eta| < 2.5$ , (i.e. Jet energy scale is defined!, detector dependent)
2. If it contains a track-jet within  $\Delta R < 0.5$
3. Some quality requirements on the track-jet.

Taggability:

$$\text{Taggability}(E_T, \eta) = \frac{\# \text{ taggable jets } (E_T, \eta)}{\# \text{ jets}(E_T, \eta)}$$

Later on, derived in 3 regions of  $z_{PV}$  (DØ)

Different parameters In CDF

# Taggability

Taggability must be derived from "generic QCD" data

Use same trigger than the signal sample, to incorporate luminosity, run number dependences...

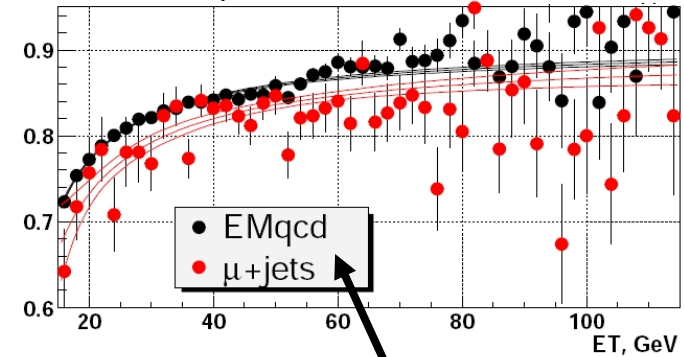
For example  $t\bar{t}$   $l+4$  jet signature, take the events passing the lepton+1 jet trigger

Signal fraction is  $\sim 10^{-4}$  : so no bias.

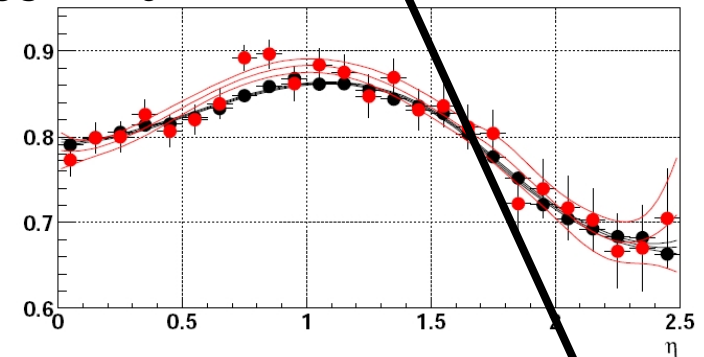
Compute taggability in bins of  $\eta$  and  $p_T$

$$\text{Taggability}(E_T, \eta) \approx k \times \text{Taggability}(E_T) \times \text{Taggability}(\eta)$$

Taggability( $E_T$ )



Taggability( $\eta$ )



## Sample dependence:

- Low MET passing EM trigger
- $\mu$  + jet + high MET sample



# Taggability and jet flavor

Taggability derived from data is valid for light flavor jets ONLY.

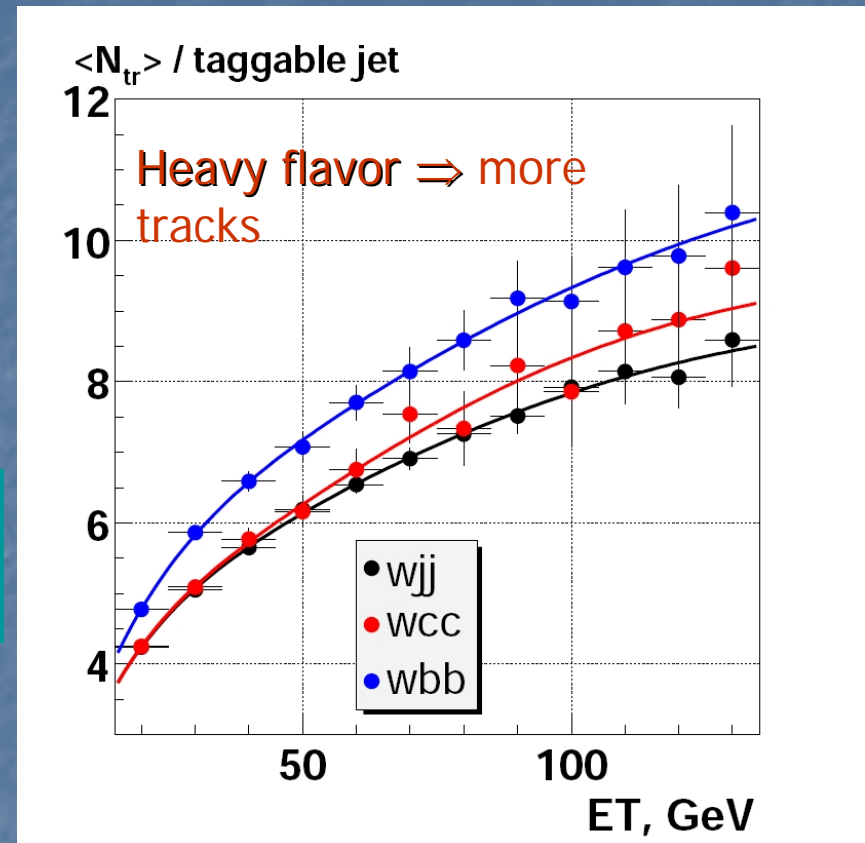
Higher taggability for heavy flavor jets

Derive correction from MC

$$C_{\text{taggability}}(\text{flavor}) = \frac{\text{Taggability}(E_T, \eta, \text{flavor}) \text{ in MC}}{\text{Taggability}(E_T, \eta, \text{light jets}) \text{ in MC}}$$

Cross check ratio of heavy-enhanced-to-light taggability in data and MC, agreement better than 2% level

$$\text{Taggability}(E_T, \eta, \text{flavor}) = C_{\text{taggability}}(\text{flavor}) \times \frac{\# \text{ taggable jets } (E_T, \eta)}{\# \text{ jets}(E_T, \eta)}$$



# b-tagging efficiency

b-tagging efficiency is defined by

$$\varepsilon^b(E_T, \eta) = \frac{\text{\# tagged jets } (E_T, \eta)}{\text{\# taggable jets } (E_T, \eta)}$$

Derive this quantity from data using a sample enhanced in heavy flavor.

Typically back-to-back dijet events with various taggers: SVT, soft muon or electron tagger

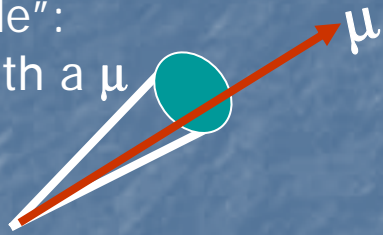
( $D\emptyset$  = a muon inside a jet with  $p_T^{\text{rel}} > 0.7$  GeV, CDF electron inside a jet)

Method introduced in  $D\emptyset$  by LEP folks...

# b-tagging efficiency from data

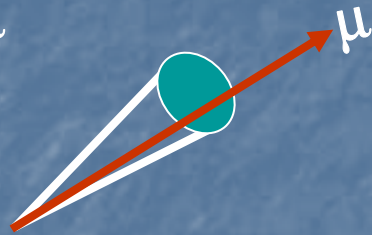
"n-sample":

$\geq 1$  jet with a  $\mu$



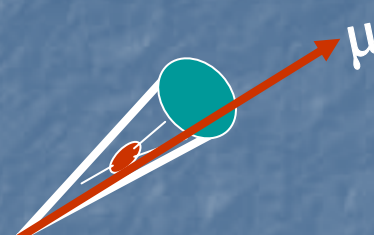
Not  $\mu$ -tagged

$n$



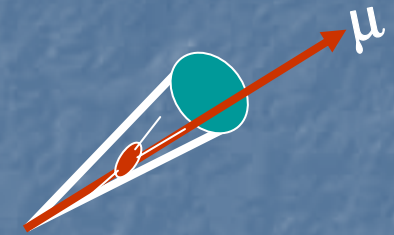
$\mu$ -tagged

$n^\mu$



Not  $\mu$ -tagged,  
SVT tagged

$n^{SVT}$



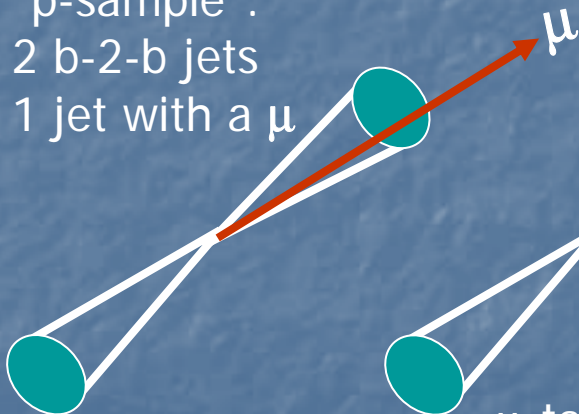
$\mu$ -tagged,  
SVT tagged

$n^{\mu,SVT}$

"p-sample":

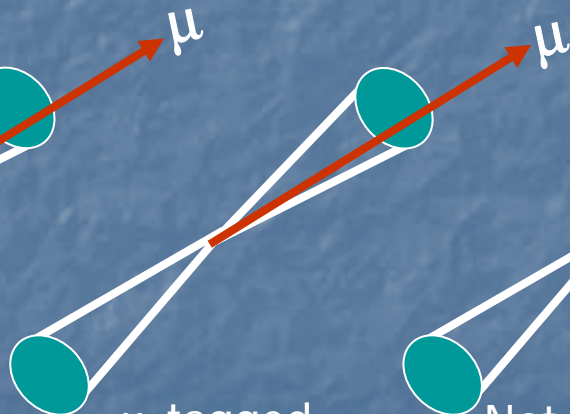
2 b-2-b jets

1 jet with a  $\mu$



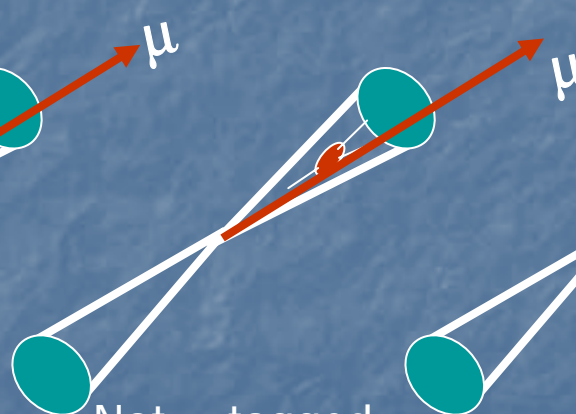
Not  $\mu$ -tagged

$p$



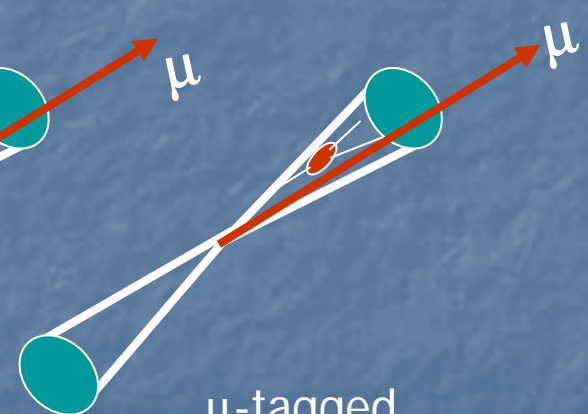
$\mu$ -tagged

$p^\mu$



Not  $\mu$ -tagged,  
SVT tagged

$p^{SVT}$



$\mu$ -tagged,  
SVT tagged

$p^{\mu,SVT}$

Solve system of 8 equations, with 8 unknowns (in bins of  $\eta$  and  $p_T$ )

$$\begin{aligned}
 n &= n_b + n_{cl} \\
 p &= p_b + p_{cl} \\
 n^\mu &= \varepsilon_b^\mu n_b + \varepsilon_{cl}^\mu n_{cl} \\
 p^\mu &= \varepsilon_b^\mu p_b + \varepsilon_{cl}^\mu p_{cl} \\
 n^{SVT} &= \varepsilon_b^{JLIP} n_b + \varepsilon_{cl}^{JLIP} n_{cl} \\
 p^{SVT} &= \beta \varepsilon_b^{JLIP} p_b + \alpha \varepsilon_{cl}^{JLIP} p_{cl} \\
 n^{\mu, SVT} &= \varepsilon_b^\mu \varepsilon_b^{JLIP} n_b + \varepsilon_{cl}^\mu \varepsilon_{cl}^{JLIP} n_{cl} \\
 p^{\mu, SVT} &= \beta \varepsilon_b^\mu \varepsilon_b^{JLIP} p_b + \alpha \varepsilon_{cl}^\mu \varepsilon_{cl}^{JLIP} p_{cl}.
 \end{aligned}$$

$n_{cl}$  ← # events that are c- or light- jets

$\varepsilon_b^\mu(E_T, \eta)$  efficiency of  $\mu$ -tagger  
 $\varepsilon_b^{SVT}(E_T, \eta)$  efficiency of 2nd tagger

b- contribution      c/light contribution

- ⇒ Extract: sample composition and efficiency of the taggers
- ⇒ Makes a number of assumptions... → systematic errors

# System 8 assumptions and systematics

- Decorrelation of the 2 taggers:

$$\epsilon^{\mu,SVT} = c \times \epsilon^{\mu} \times \epsilon^{SVT}, \text{ assume } c=1 \text{ (MC gives } c=1.01\pm 0.01)$$

- Assume that the  $\mu$ -tagger has same efficiency for c- and light-jets, ok because  $p_T^{rel}$  has similar shape for c- and light-jets at Tevatron energy.
  - Compare  $p_T^{rel}$  templates from several generators
- Assume that c- and light-jet backgrounds can be lumped together, this is characterised by a factor  $\alpha$  (varied for systematics)
- Solve the system for various values of  $p_T^{rel}$  cut 0.3 - 1.5 GeV.
- $\beta \sim 1$  takes into account correlations b/w p and n samples (varied for systematics)

# b-tagging efficiency

From data we can only extract b-tagging efficiency for muonic b-jets

$$\varepsilon^{b \rightarrow \mu, \text{data}}(E_T, \eta)$$

We need the b-tagging efficiency for "all kinds of b-jets"

$$\varepsilon^b(E_T, \eta)$$

$$P_{\text{tag}}^b(E_T, \eta) =$$

$$\frac{\varepsilon^{b, \text{MC}}(E_T, \eta)}{\varepsilon^{b \rightarrow \mu, \text{MC}}(E_T, \eta)} \times \varepsilon^{b \rightarrow \mu, \text{data}}(E_T, \eta) \times \text{Taggability}(E_T, \eta) \times C_{\text{taggability}}(b)$$

Transform semi-muonic b-tag efficiency into inclusive one

# Mistag rate

Extract from data as much as possible  
 Similar approach in CDF and DØ

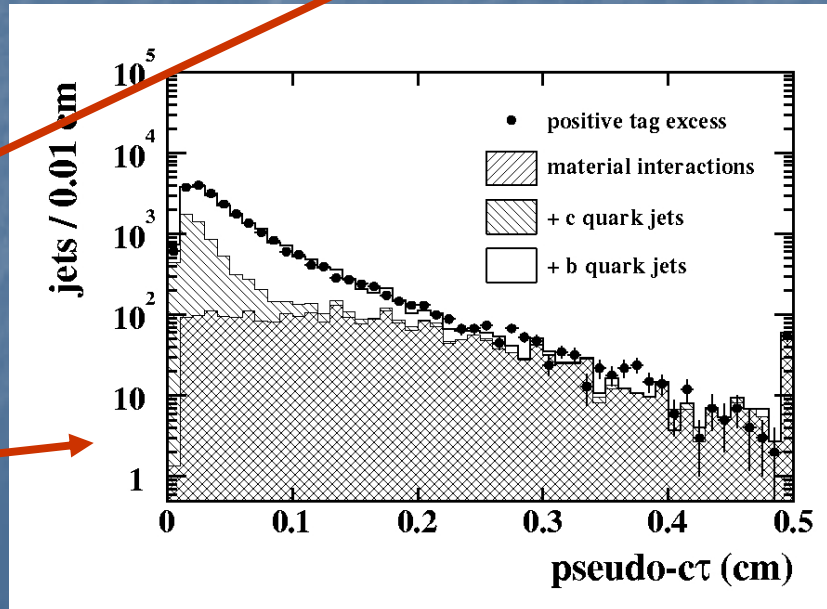
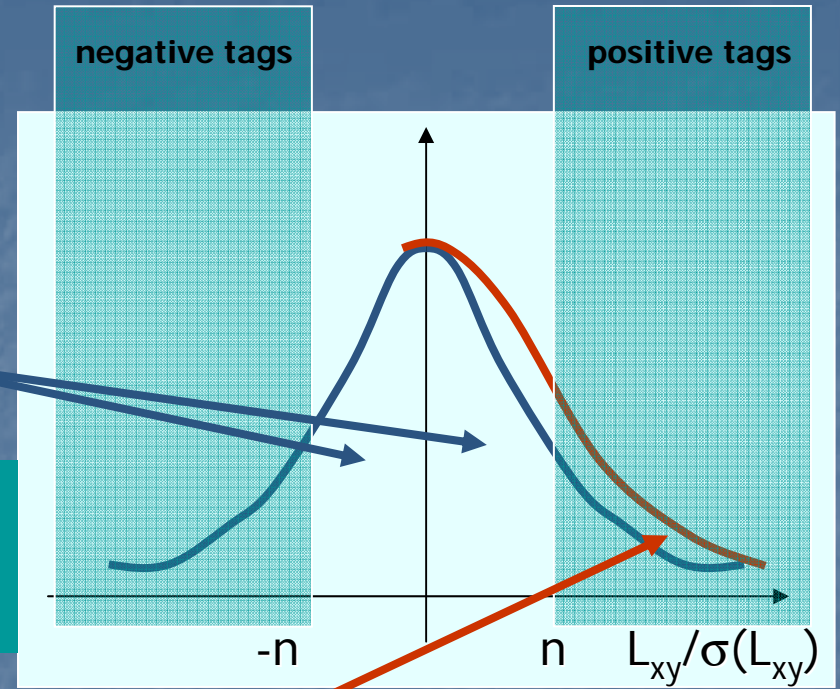
The number of negative tags gives  
 information on resolution effects

$$\epsilon_{\text{data}}(E_T, \eta) = \frac{\text{\# negatively tagged jets } (E_T, \eta)}{\text{\# taggable jets } (E_T, \eta)}$$

Must take into account:

- 1) Long-lived particles in light jets are not completely removed by  $V^0$  filter: contribute to positive tags in light jets
- 2) Contamination of negative tags data by heavy flavor (2% b-jets and ~4% c-jets)

CDF fits contributions to observed pseudo- $c\tau$



$\varepsilon_{\text{data}}(E_T, \eta)$  extracted in data passing e+jet trigger, with MissingET < 10 GeV

Validation:

1. Alternative parametrization derived from single electron trigger
2. Compare predicted and observed number of negative tags in high MissingET region

Correct for long-lived particles in light-jet sample:

$$SF_{\parallel} = \text{\#negative tags} / \text{\#positive tags} \quad \text{in light-flavor QCD Monte Carlo}$$

Correct for the fraction of heavy flavor in the low MissingET electron sample

$$SF_{\text{hf}} = \text{\#positive tag from light flavor} / \text{\# positive tag from all flavors}$$

$$\varepsilon^{\text{light}}(E_T, \eta) = \varepsilon_{\text{data}}(E_T, \eta) \times SF_{\text{hf}} \times SF_{\parallel}$$

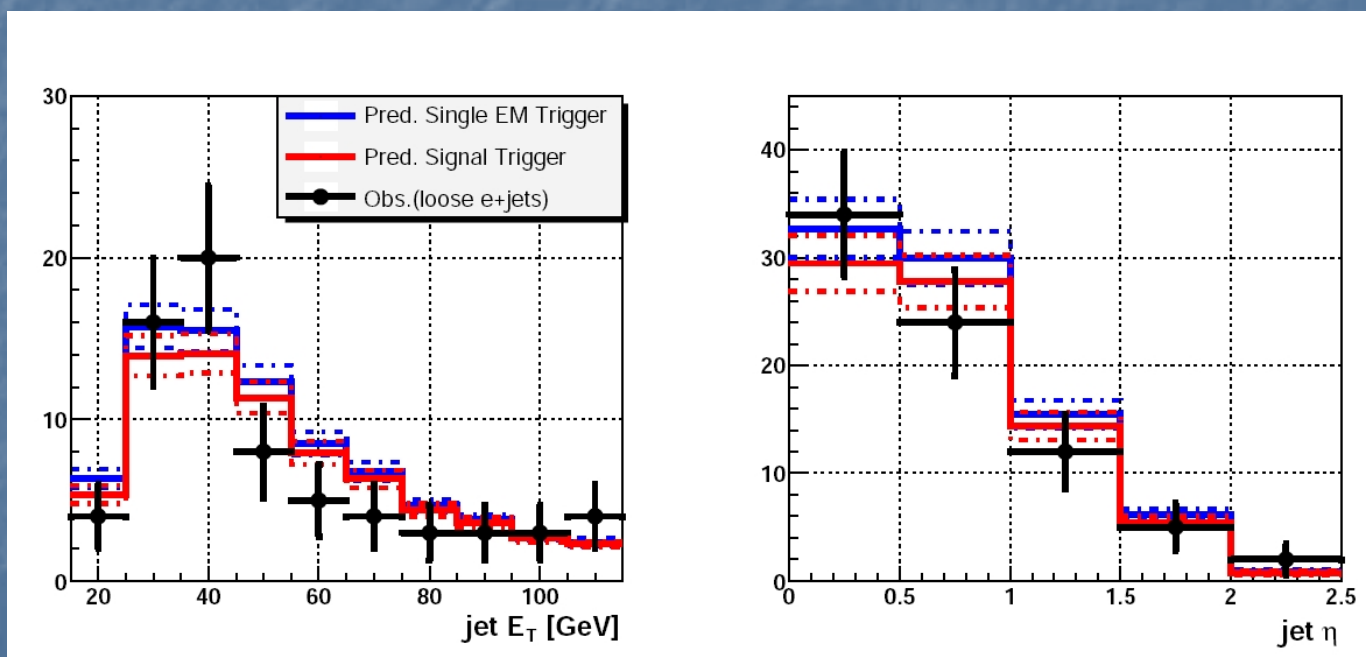


# Negative tag rate validation

Alternative parametrizations derived:

- from single electron trigger
- (instead of e+jets)

Compare predicted and observed number of negative tags in high MissingET region



# Systematic uncertainties

## I. Taggability

1. **Statistical error on parametrization from data**
2. Variation on the parametrization by changing sample
3. Difference b/w predicted and observed # taggable jets at high  $N_{jet}$
4. Flavor dependence of taggability: MC dependence

## II. b-tagging efficiency

1. **Statistical error on semi-muonic b-tagging parametrization from data**
2. System-8 assumptions
3. Ratio of semi-muonic to inclusive b-tagging efficiency in MC (statistical+sample dependence)

## III. c-tagging efficiencies

## IV. Mistag rate

1. Negative tag rate, data statistics
2. Negative tag rate, sample dependence
3. Heavy flavor contamination
4. Negative to positive tag ratio for light flavor jets