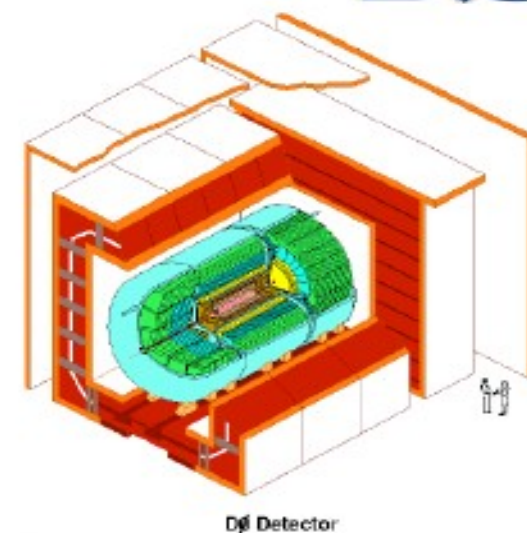
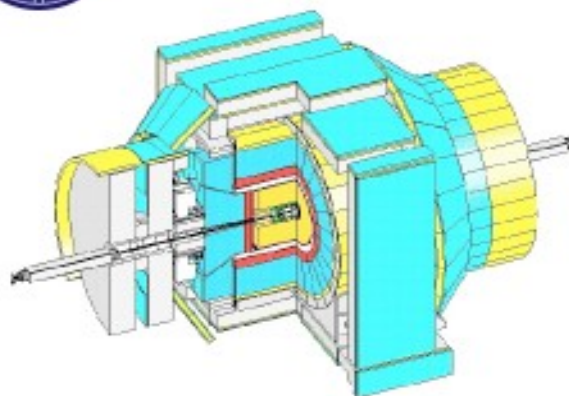




Introduction



B-Tagging is critical to physics at the Tevatron

Top ($t \rightarrow Wb$, *single-top*)
SUSY (stop, sbottom, ...)
Higgs ($h \rightarrow bb$)

We've learned a lot...

Smart (multivariate) b/c reconstruction
Accurate measurement of efficiency / fake rate
Optimal *use* of info in analyses



Disclaimers and Qualifications

I'm on D0

- Biased towards our methods
- But CDF is similar, and we do share techniques

b-ID convener, for past 2 years

Now Higgs convener

- Single-top has slightly different needs...

I like searches!

- Other issues may dominate for precision measurements (top cross-section, etc.)

Tevatron (p-pbar) is an extremely difficult environment

- QCD background!
 $\sigma(Z \rightarrow bb) / \sigma(bb) = \sim 0.05$ (under peak)!
- Personpower-limited
 ~ 150 FTE / experiment!

b/c Tagging Techniques

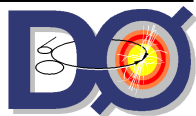
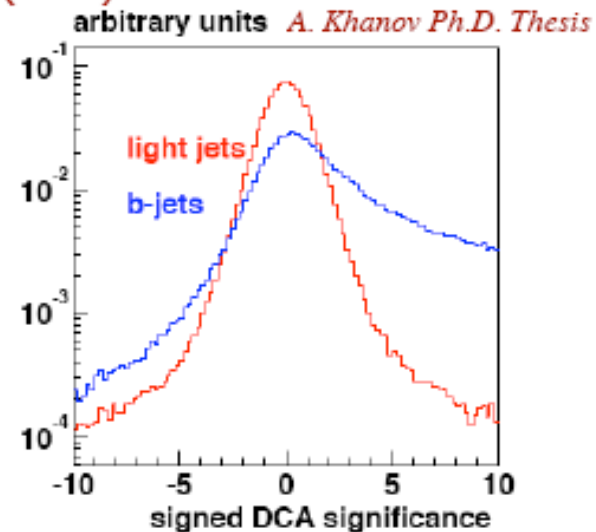
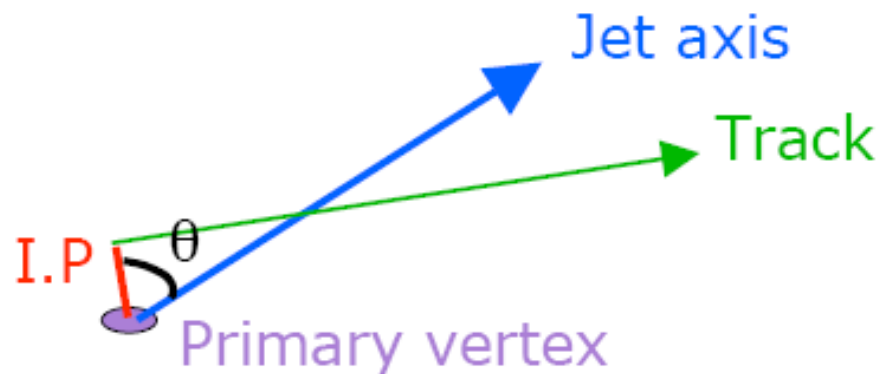
Tagging Algorithms



- DØ b-tagging group: several groups worked on development and evaluation of different algorithms:

1. Counting Signed Impact Parameter:

- simple algorithm and robust design
- Based on Impact Parameter Significance: $S(IP) = IP/s(IP)$
- IP is a signed quantity w.r.t to jet axis (so is $S(IP)$):
 - positive if $\theta < \pi/2$, negative if $\theta > \pi/2$
- **Requirements to tag a jet positively (negatively):**
 - at least 2 tracks with $S(IP) > 3$ (< -3)
 - or at least 3 tracks with $S(IP) > 2$ (< -2)



Tagging Algorithms

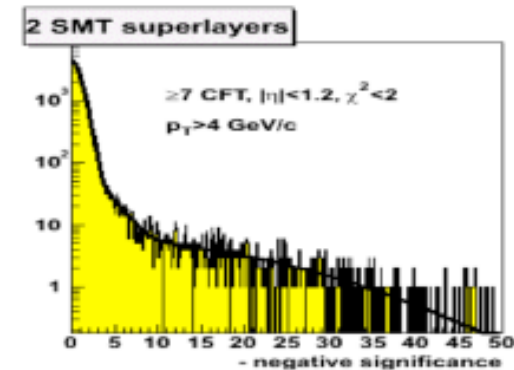


- DØ b-tagging group: several groups worked on development and evaluation of different algorithms:

1. Counting Signed Impact Parameter:

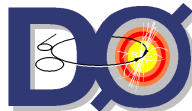
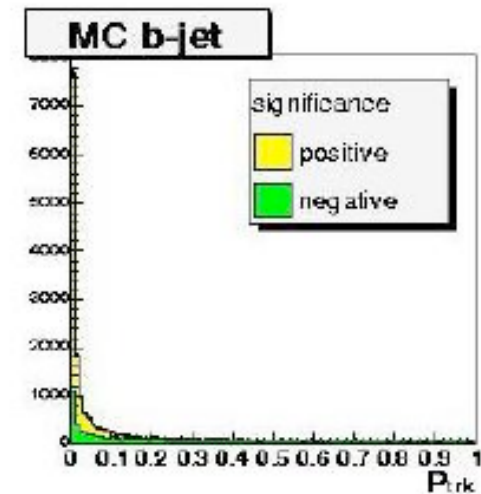
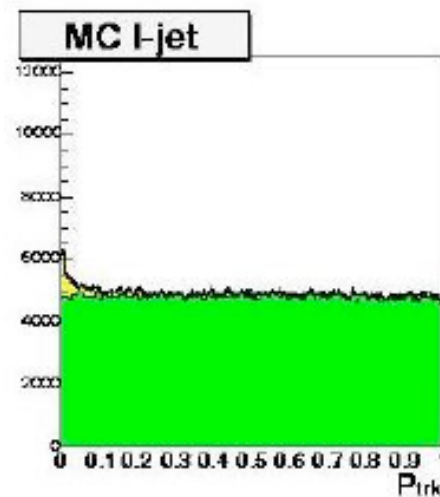
2. Jet Lifetime Impact Parameter:

- Fit a resolution function (R) on the negative S(IP) distribution. Assume these tracks to **originate from primary vertex**.
 - Several track categories defined
- Use then R to define a **probability** for a track to originate from the primary vertex



S. Greder Ph.D. Thesis

$$P_{\text{trace}}(S_{IP}) = \frac{\int_{-50}^{-|S_{IP}|} \mathcal{R}(s) ds}{\int_{-50}^0 \mathcal{R}(s) ds} \quad (4.6)$$



Tagging Algorithms

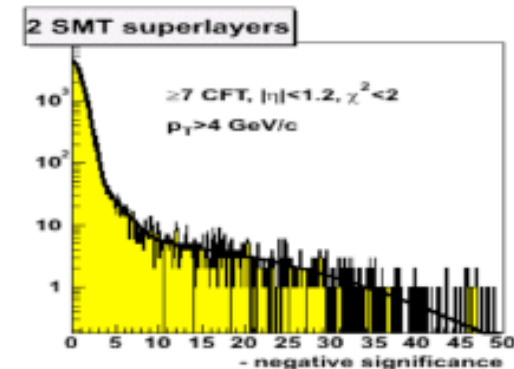


- DØ b-tagging group: several groups worked on development and evaluation of different algorithms:

1. Counting Signed Impact Parameter:

2. Jet Lifetime Impact Parameter:

- Fit a resolution function (R) on the negative S(IP) distribution. Assume these tracks to **originate from primary vertex**.
 - Several track categories defined
- Use then R to define a **probability** for a track to originate from the primary vertex
- Combine then each track's probability to compute the probability for N tracks to originate from primary vertex : P_{jet} the jet lifetime probability



S. Greder Ph.D. Thesis

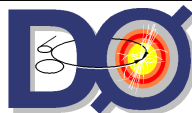
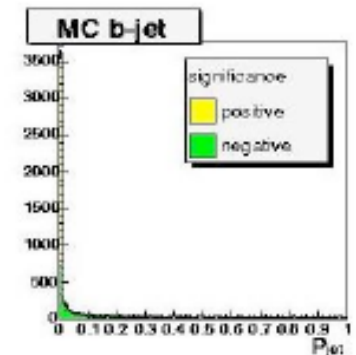
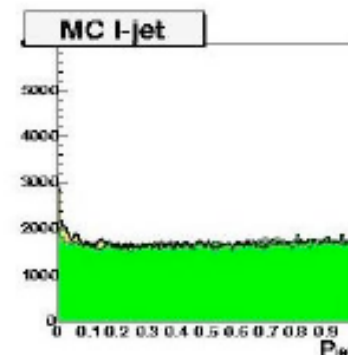
$$\mathcal{P}_{jet}^{\pm} = \Pi^{\pm} \times \sum_{j=0}^{N_{traces}^{\pm}-1} \frac{(-\log \Pi^{\pm})^j}{j!} \quad \text{ou} : \quad \Pi^{\pm} = \prod_{i=1}^{N_{traces}^{\pm}} \mathcal{P}_{traces}(S_{IP}^{IP>0}) \quad (4.7)$$

Requirements to tag a jet positively :

- $P_{jet} (+) < \text{cut}$

Requirements to tag a jet negatively :

- $P_{jet} (-) < \text{cut}$



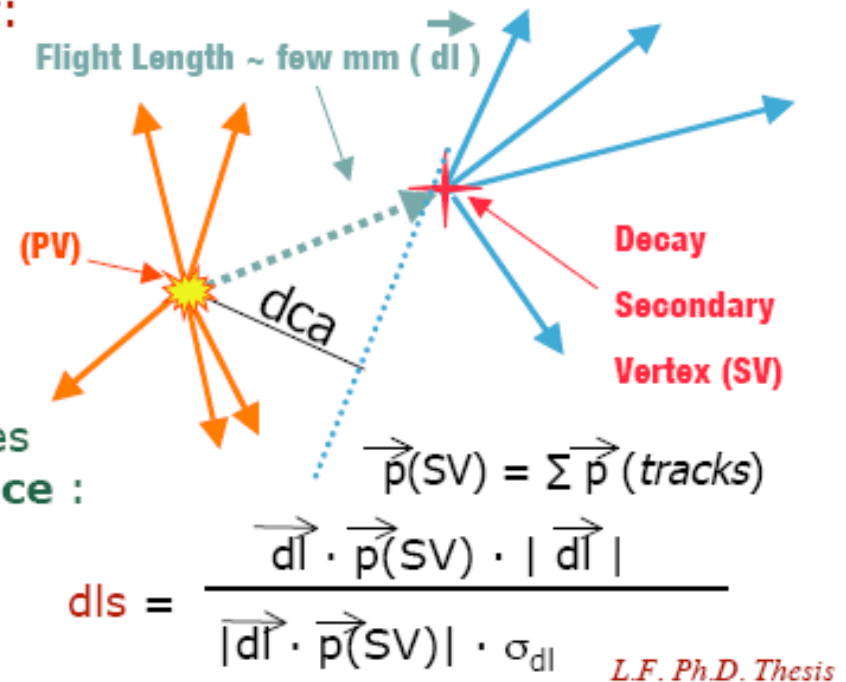
Tagging Algorithms



- DØ b-tagging group: several groups worked on development and evaluation of different algorithms:

1. Counting Signed Impact Parameter:
2. Jet Lifetime Impact Parameter:
3. Secondary Vertex Tagger:

- Build up track-based jets and fit their tracks to a secondary vertex
- Select tracks with **high IP** to build secondary vertices
- Tracks with high IP will form vertices with high **decay length significance** :
 $S(L_{xy}) = L_{xy}/s(L_{xy})$.
- Jet are then tagged by requiring a **dR < 0.4 match** between the jet axis and the SV.

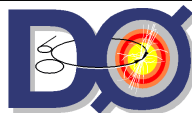
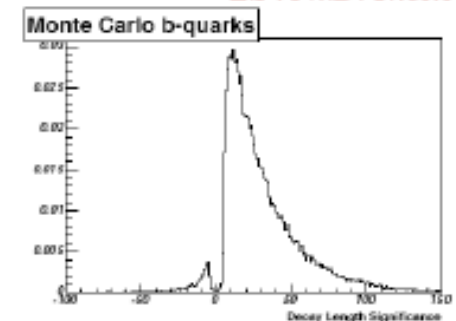
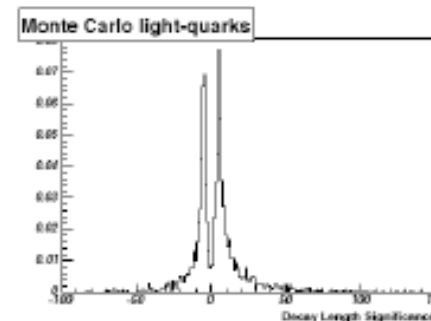


Requirements to tag a jet positively :

- match a secondary vertex with $S(L_{xy}) > \text{cut}$

Requirements to tag a jet negatively :

- match a secondary vertex with $S(L_{xy}) < -\text{cut}$



Tagging Algorithms



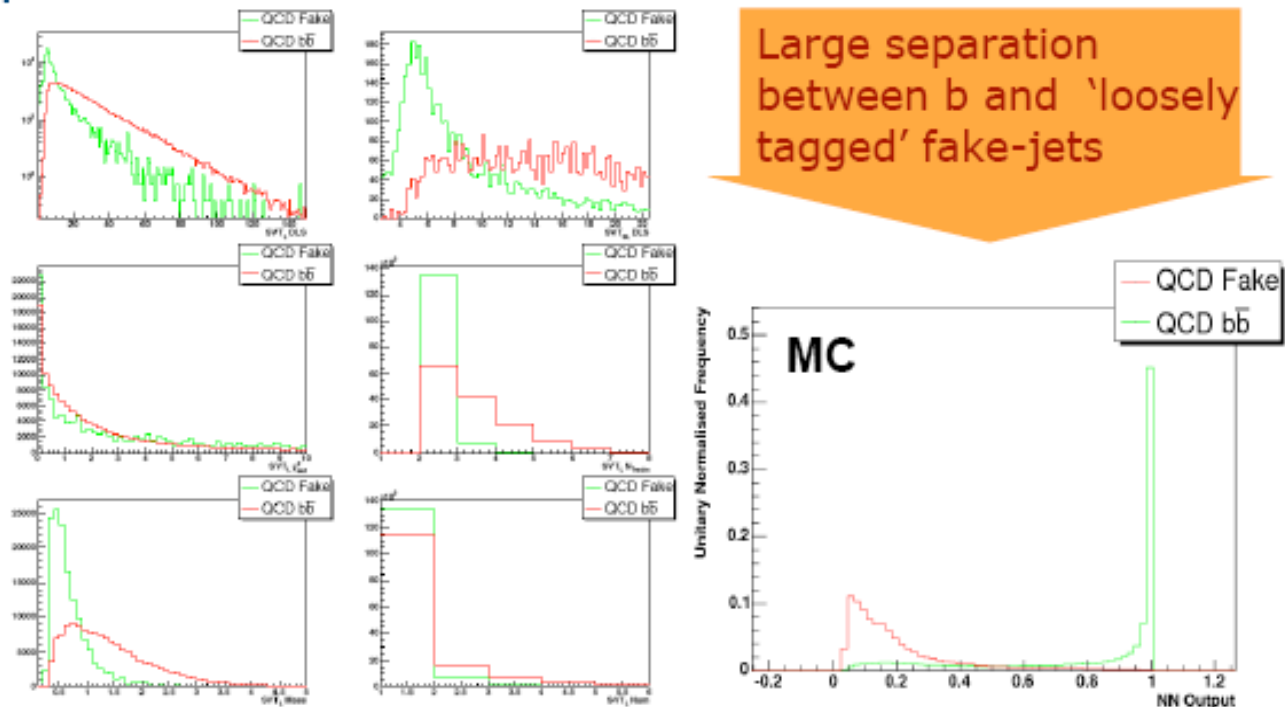
- DØ b-tagging group: several groups worked on development and evaluation of different algorithms:
 1. Counting Signed Impact Parameter:
 2. Jet Lifetime Impact Parameter:
 3. Secondary Vertex Tagger:
- A NN was developed using these different taggers to enhance b-tag performances



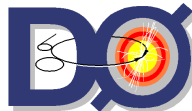
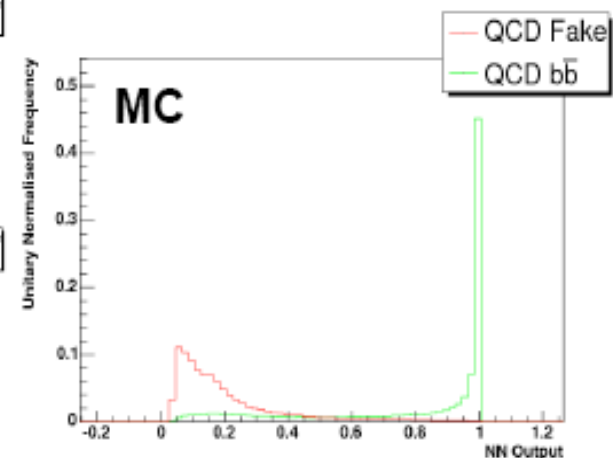
T. Scanlon Ph.D. Thesis

NN variables

Variable
SVT _{SL} DLS
CSIP Comb
JLIP Prob
SVT _{SL} χ^2_{dof}
SVT _L N_{Tracks}
SVT _{SL} Mass
SVT _{SL} Num
JLIP $Prob_{red}$
SVT _{SL} dR



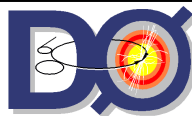
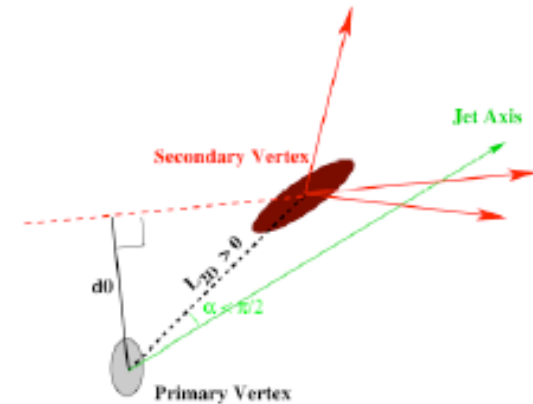
Large separation between b and 'loosely tagged' fake-jets





Tagging Algorithms

- Used for 1st observation of top quark
- Two successive attempts:
 - here shown TIGHT SecVtx configuration:
- Pass 1:
 - Loose track selection ($p_T > 0.5$ GeV)
 - Require > 2 displaced tracks ($|S(d_0)| > 2$)
 - Attempt 3D Secondary Vertex reconstruction from displaced tracks (prune incompatible tracks iteratively until converge)
- Pass 2:
 - Tighter track selection but require only 2 displaced tracks ($S(d_0) > 3.5$, one with $p_T > 1.5$ GeV, other with $p_T > 1$ GeV)
 - Attempt Secondary Vertex reconstruction from two tracks
- Tag if found a SV. and $L2D/\sigma(L2D) > 7.5$

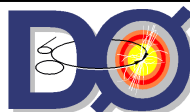
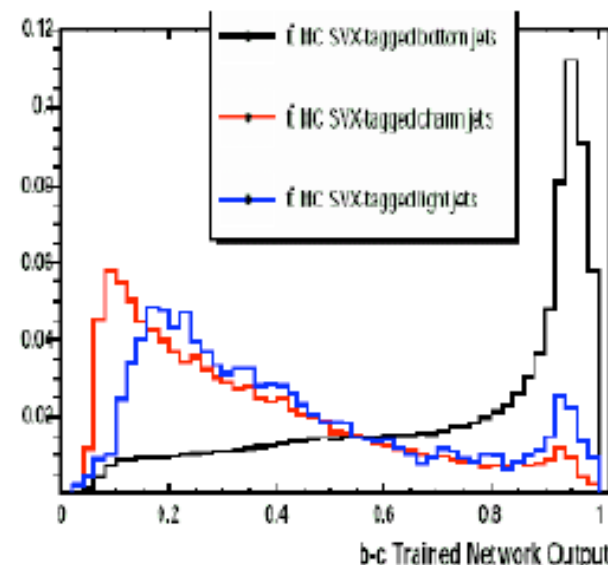
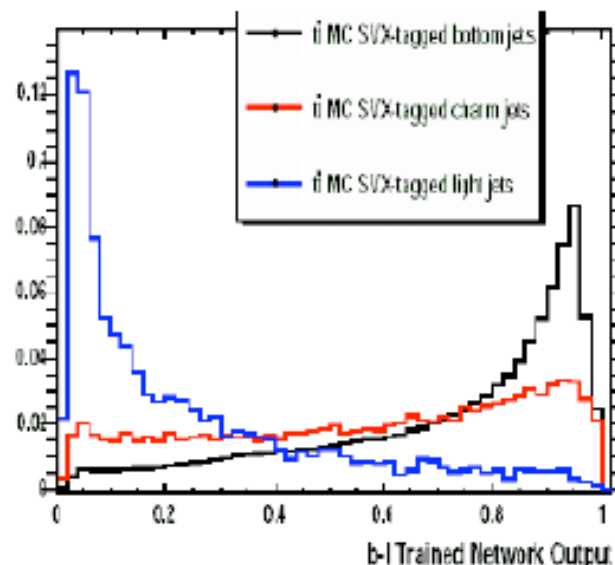




Tagging Algorithms

- Start with a SecVtx tag and filter out c and light jets with NN
- 16 variables (8 relative to SecVtx):

SECVTX variable	SECVTX independent variable
• Number of tracks in SECVTX	• Number of good tracks
• Fit χ^2	• Jet Probability (JetProb)
• Transverse decay length (L_{xy})	• Reconstructed mass of pass 1 tracks
• L_{xy} significance ($L_{xy}/\sigma_{L_{xy}}$)	• Reconstructed mass of pass 2 tracks
• Pseudo- $c\tau$ ($L_{xy} \times M_{SECVTX}/p_T^{SECVTX}$)	• Number of pass 1 tracks
• Vertex Mass ($\sqrt{(\sum p_{vtx})^2 - (\sum p_{vtx})^2}$)	• Number of pass 2 tracks
• $p_T^{vtx}/(\sum_{good\ tracks} p_T)$	• $\sum_{Pass1\ track} p_T/p_T^{jet}$
• Vertex pass number (pass 1 or 2)	• $\sum_{Pass2\ track} p_T/p_T^{jet}$

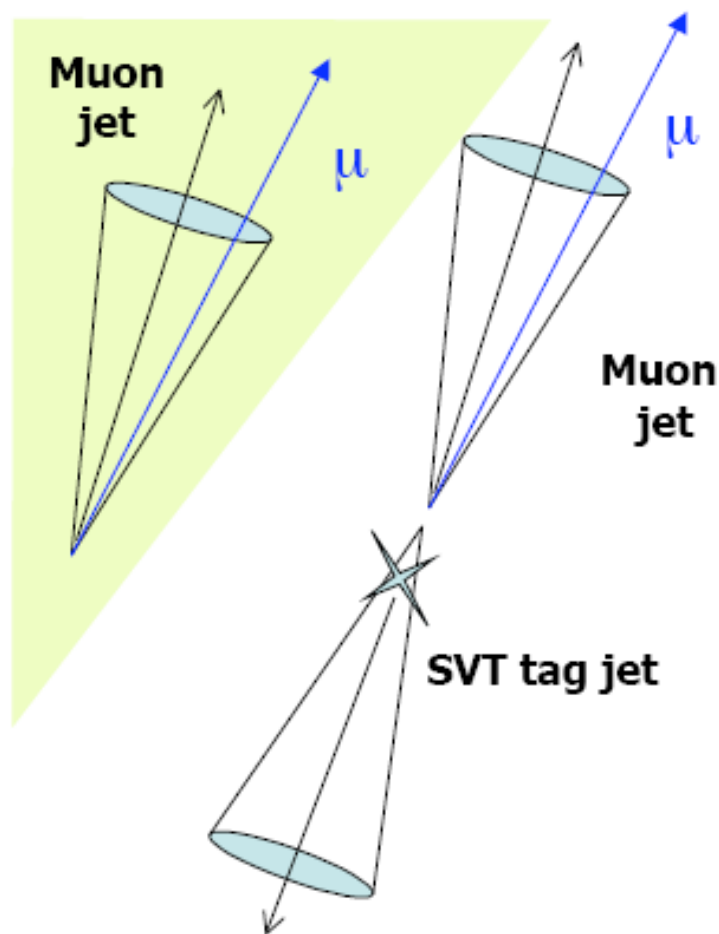


Measuring b/c Tagging Efficiencies *in data*

b-tagging efficiency from data



- System 8: System of **8** equation and **8** unknowns
 - 2 samples
 - Muon Jets (n)
 - MJ + opposite lifetime tag Jet (p)
 - 2 different b and light content
 - $n_b \neq p_b$
 - 2 uncorrelated tagging algorithms
 - Track based
 - Muon Tag ($p_T^{\text{rel}} > 0.8 \text{ GeV}$)
- 2 taggers are applied separately and together to the two different samples
- Correlation among the two taggers is derived from simulation
 - only MC input
- **System 8 is analytically solvable!**
- High statistics allow to solve the system in bins of jet p_T and η
- DØ only b -tagging efficiency estimation algorithm





Efficiency measurement

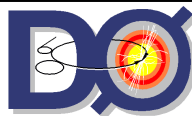
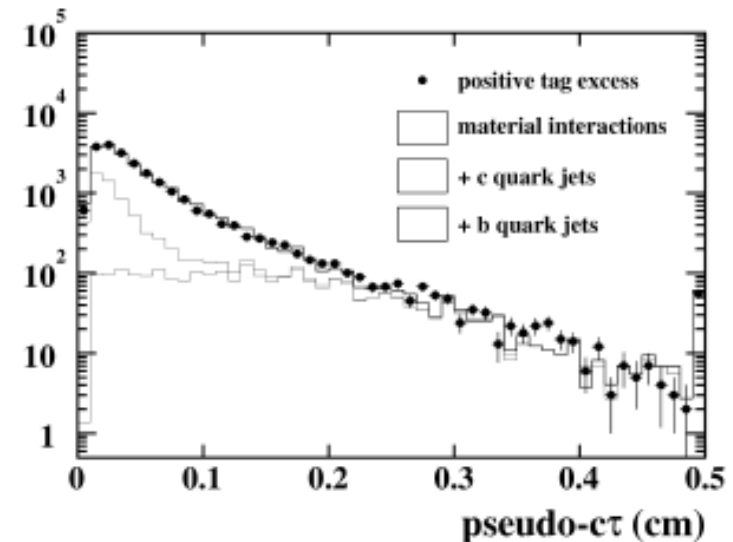
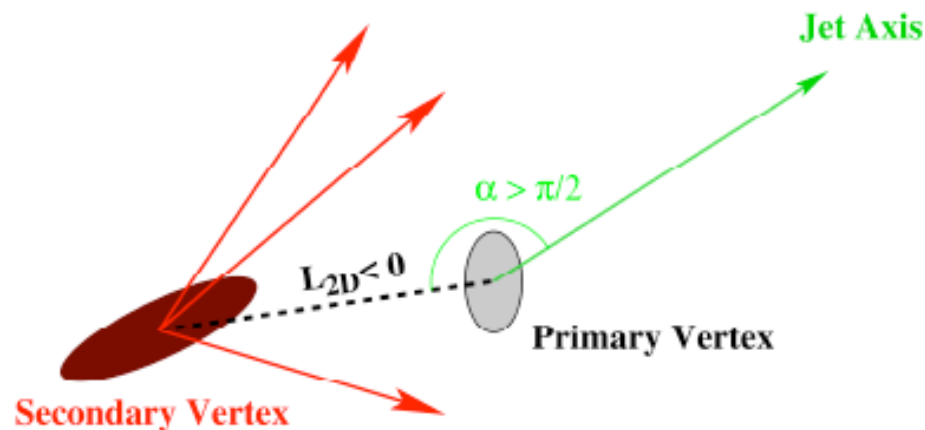
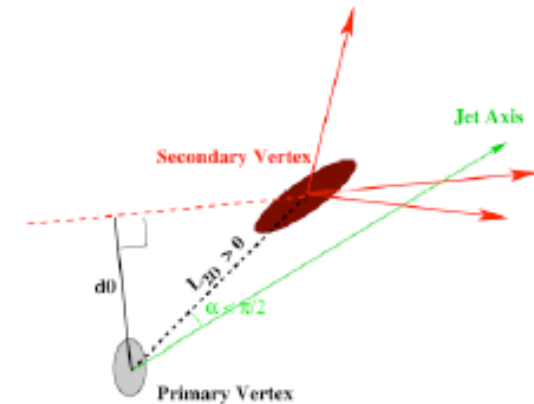
- Efficiency measured in data
 - «Scale Factor» = $\text{eff}(\text{data})/\text{eff}(\text{MC})$ is a constant
 - MC describes well enough non-uniformity (ET, eta, phi, # tracks) -> taken as systematic error
- Use di-jet sample triggered by 8 GeV electron or muon
- Opposite jet b-tag used to purify lepton-jet HF content
 - Electron: use conversion to estimate fraction of light jets
 - Muon: use p_T (relative to jet axis)
- Efficiency in data is $\sim 10\%$ too low. Estimated at 7% level. Scale Factor = 0.909 ± 0.06 (as of last year)
- Scale Factor NOT efficiency applied for physics analysis





Mistag rate from data

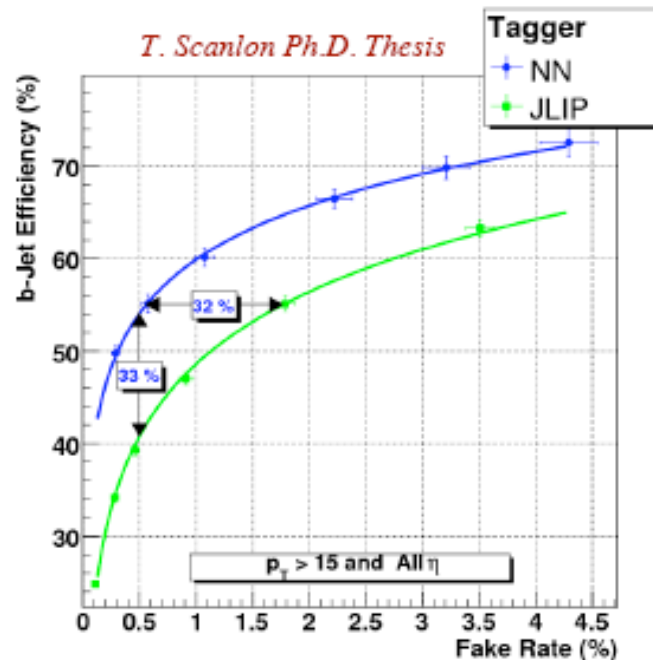
- «Negative» tag rate parametrized in several jet samples, as a function of:
 - Jet ET, eta, phi, #tracks in jet, Sum(ET)
 - Correct for asymmetry between negative and positive tag rate
 - decay in flight plus material interaction
 - Estimate this asymmetry with c-tau fit in jet samples. Pos fake = $1.37 \pm 0.15 \cdot$ Neg Fake



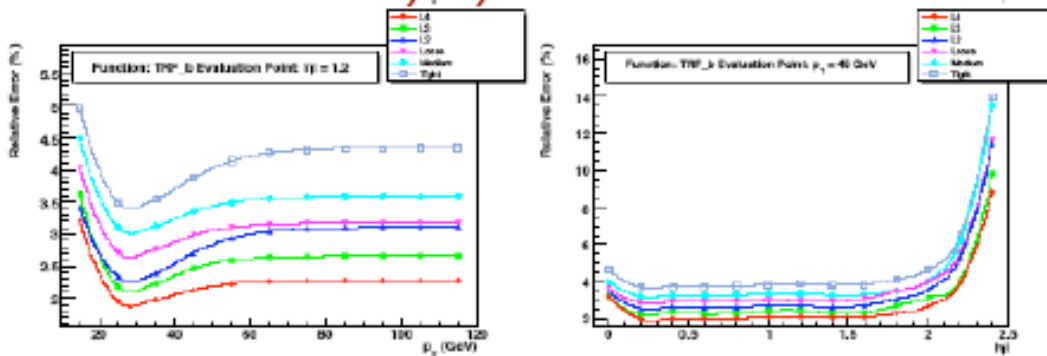
Tag Rate Functions



- Taggers are optimized for several working points
 - Efficiencies, scale factors, systematic uncertainties calculated only for these settings
- Different sources of systematic errors:
 - System 8 correlation functions
 - efficiency is recalculated after varying correlation factors
 - Parameterization error
 - correlation between p_T and η TRF parameterizations
 - MC sample dependencies

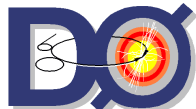


NN efficiency systematic errors



Analysis Strategies

- Use Tag Rat Functions
 - derived from data
 - known systematics
- Use Scale Factors
 - b-tagging on MC
 - allows MC to predict unknown source of systematics
 - I.e. hidden taggability dependency

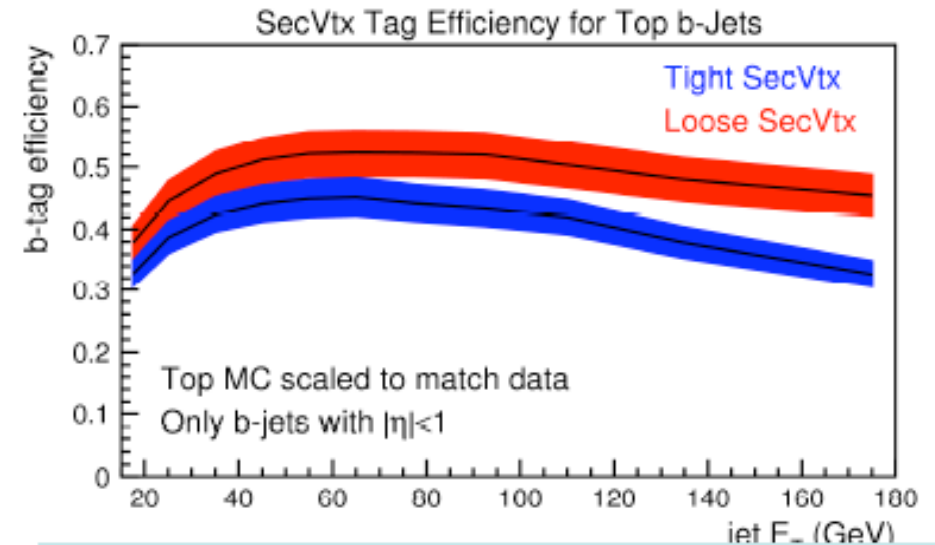
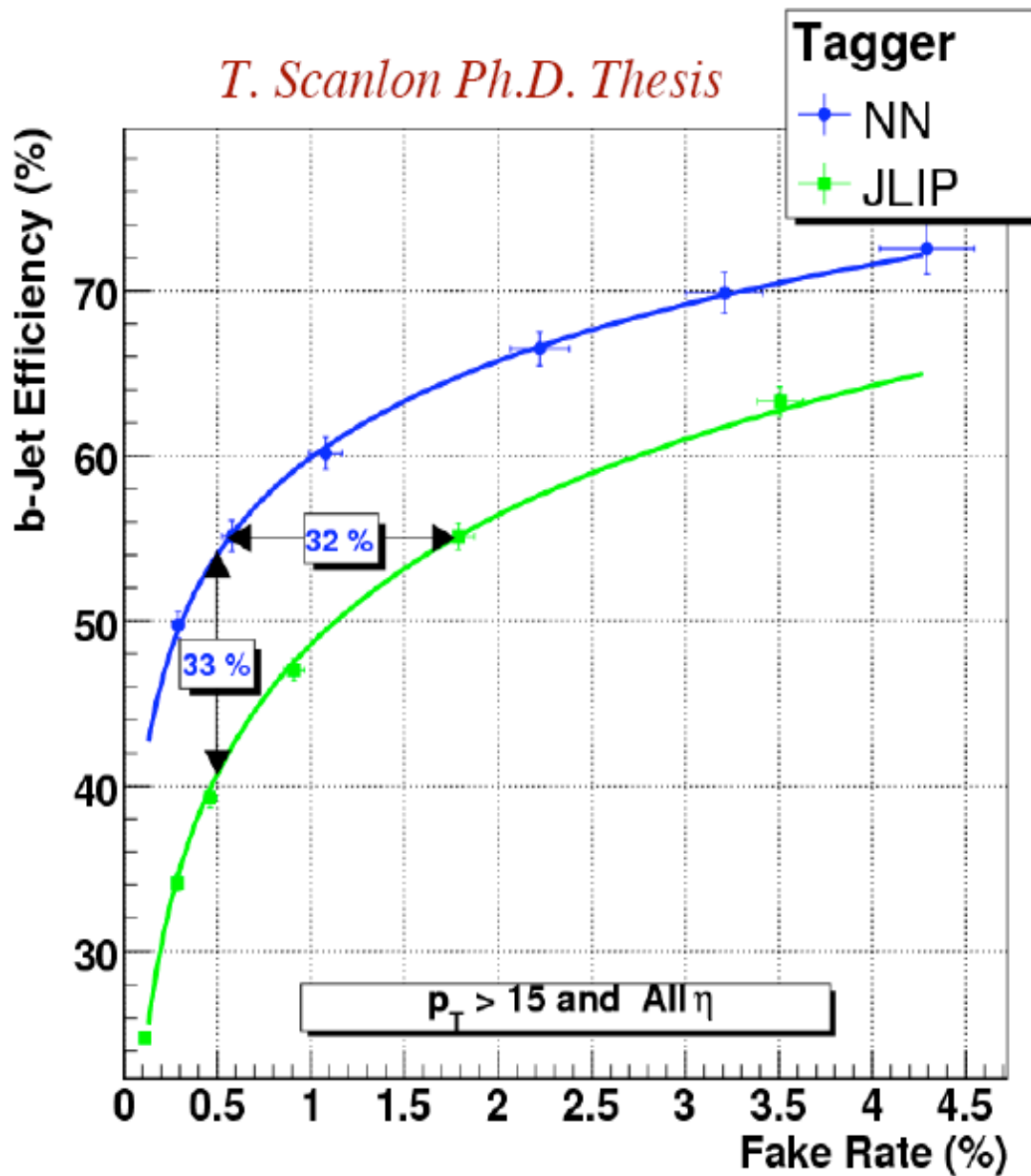




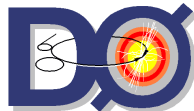
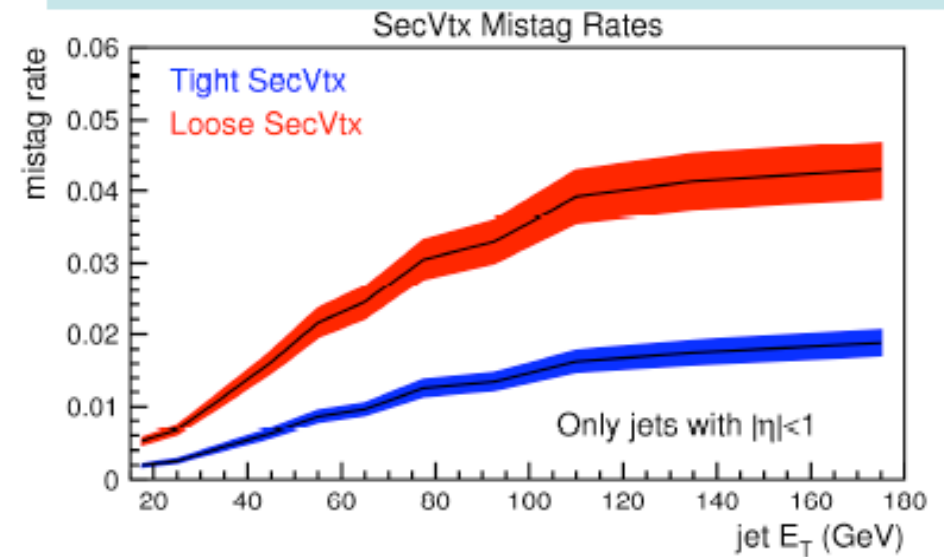
Performance



T. Scanlon Ph.D. Thesis



Eff Scale Factor = 0.909 ± 0.06 for the tight



Use of b-tagging in physics analyses

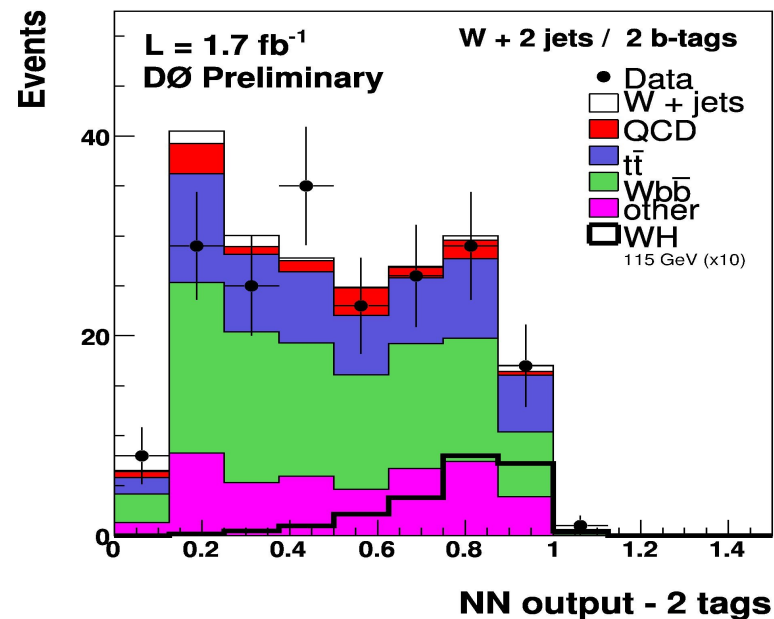
Single-tight / Double-loose

WH(->bb) as an example

- Note: 2 b's in most final states!

Base:

Find optimal double b-tag working point (usually very loose)



Single-tight / Double-loose

WH(->bb) as an example

- Note: 2 b's in most final states!

Base:

Find optimal double b-tag working point (usually very loose)

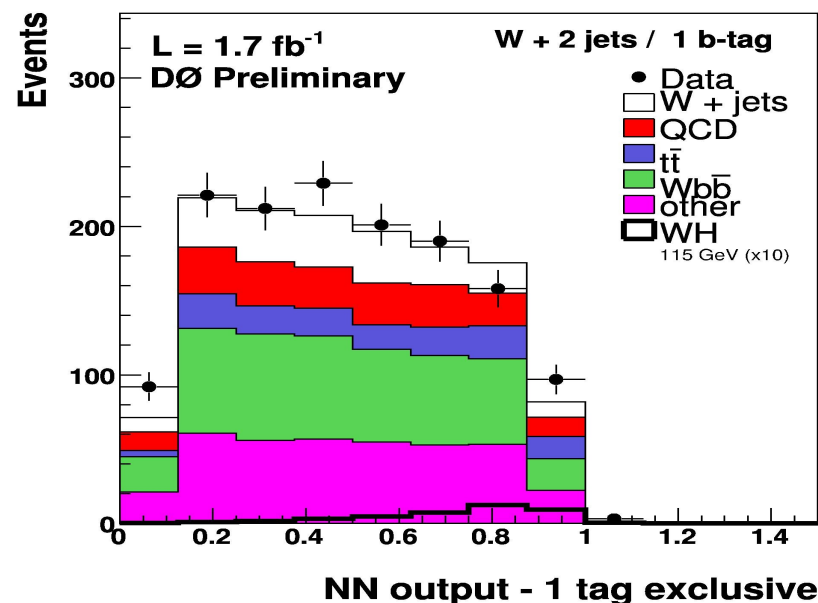
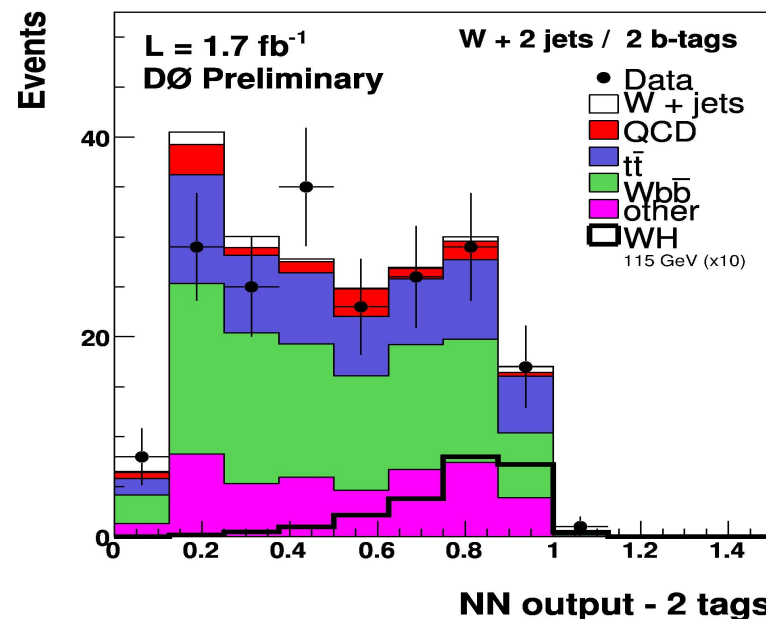
Better:

Add optimal (orthogonal) single b-tag working point

- Usually much tighter
- Careful: adding single tag can change the optimal double tag working point!

Gain in performance:

~10% in sensitivity compared to one working point



Two Working Points

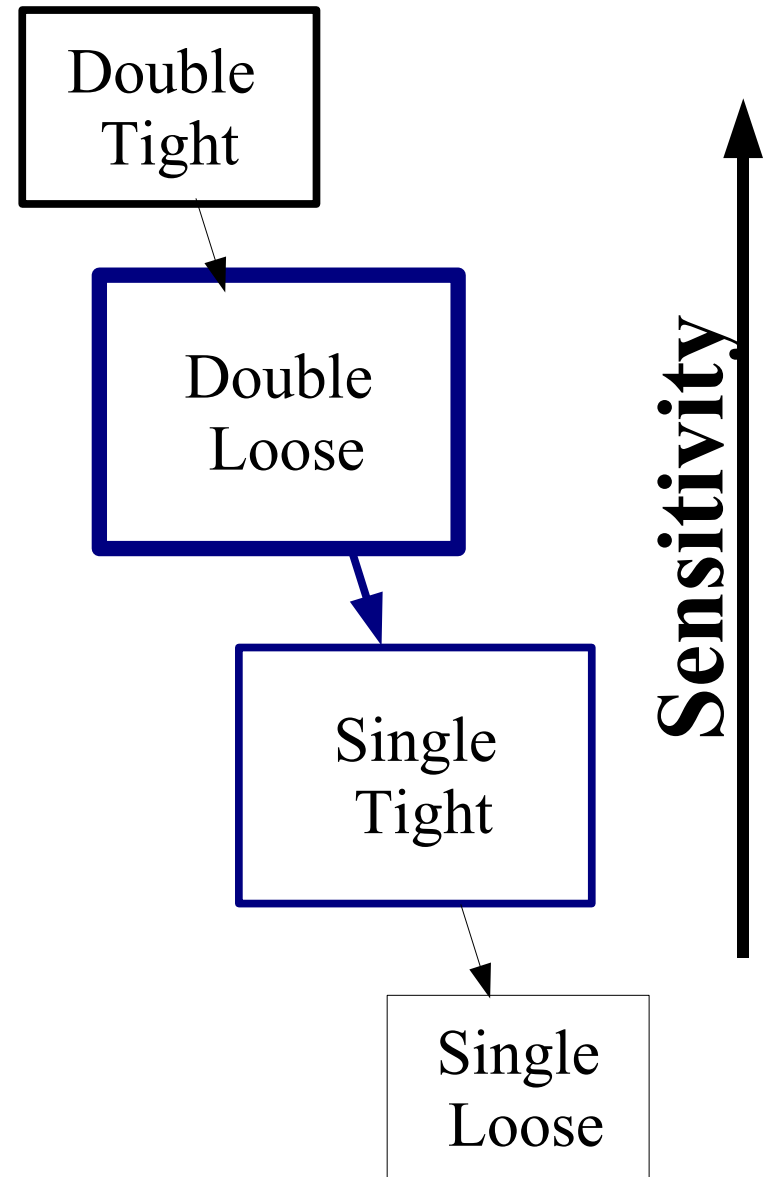
Better still:

Add double-tight working point
And single-loose working point

(Keep selections orthogonal!)

Gain in performance:

~2% in sensitivity compared to two
working points



Many Working Points

Even Better Still:

Use 3,4,... 12 working points

Including *untagged* there are 169 boxes (1 and 2 tags)

Order in terms of sensitivity (and keep orthogonal)

Gain in performance:
~10% in sensitivity compared to two working points

B-tagging discriminant

Define tagging variables for jet 1:	$l_1 = TRF_l^x - TRF_l^{x+1}$	where jet 1 is between operating point x and x+1
	$c_1 = TRF_c^x - TRF_c^{x+1}$	
	$b_1 = TRF_b^x - TRF_b^{x+1}$	Likewise for jet 2

Then build our discriminant: $D = \frac{S}{S+B}$ $S = ME_{WH} b_1 b_2$

$$\begin{aligned}
 B = & k_{wgg, ll} ME_{wgg} l_1 l_2 + k_{wgg, cl} ME_{wgg} c_1 l_2 + k_{wgg, cl} ME_{wgg} l_1 c_2 + \dots \\
 & + k_{schan, bb} ME_{schan} b_1 b_2 + k_{tchan, bl} ME_{tchan, 1} b_1 l_2 + k_{tchan, bl} ME_{tchan, 2} b_2 l_1 + \dots \\
 & + k_{ww, ll} ME_{ww} l_1 l_2 + k_{ww, cl} ME_{ww} c_1 l_2 + k_{ww, cl} l_1 c_2 + \dots \\
 & + k_{wz, ll} ME_{wz, ll} l_1 l_2 + k_{wz, cc} ME_{wz} c_1 c_2 + k_{wz, bb} b_1 b_2 + \dots
 \end{aligned}$$

$$\sum k = 1$$

10/17/07

Alan Magerkurth

3/13

Infinite Working Points

Use the output of the NN tagger

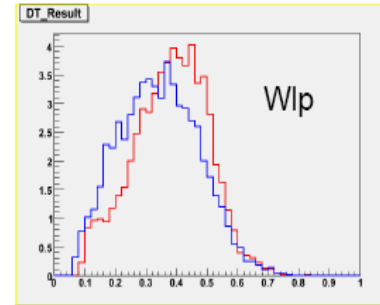
Danger: how to account for data/MC scale factors?

Under development

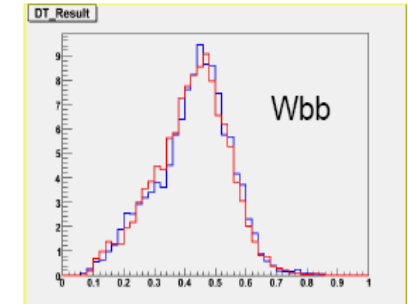
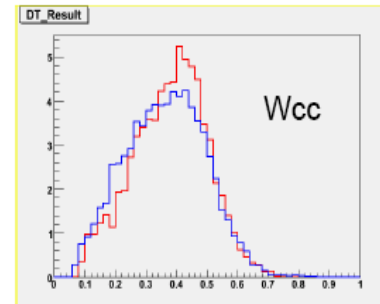
Gain in performance:

~0% in sensitivity compared to 12 working points

But can be added much more simply as an input to another NN...



- Red is DT output with binary NNTag variable
- Blue is DT output with continuous NN output



John BackusMayes

Soft-lepton Tagging



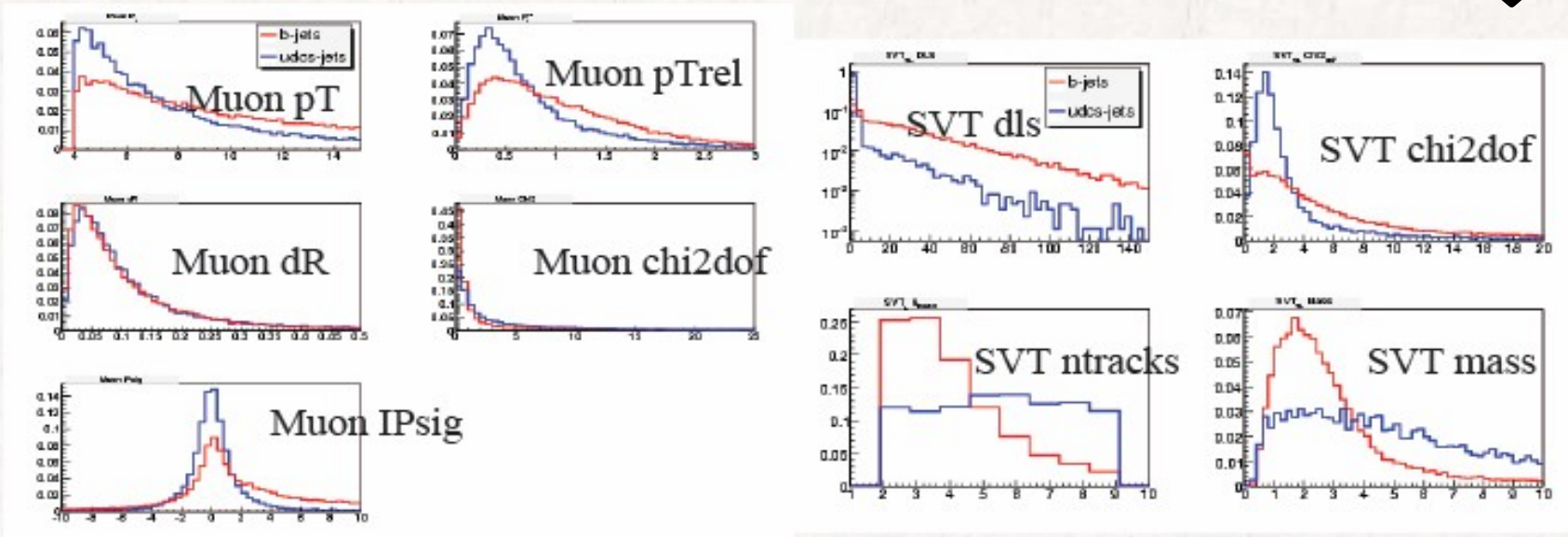
Neural Network



Input variables

- muon: P_t^{rel} , IP significance, ΔR , p_T , χ^2_{dof}
- SVT: $\text{SVT}_{\text{SL}}\text{DLS}$, $\text{SVT}_{\text{SL}}\chi^2_{\text{dof}}$, $\text{SVT}_{\text{L}}N_{\text{tracks}}$, $\text{SVT}_{\text{SL}}\text{Mass}$
- SVT variables involved with muon track only

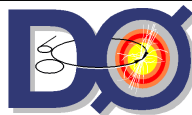
*Muon-only
so far!*



09/27/07 b-id Meeting

Hwidong Yoo

Slide 6



tau Tagging

Use a tau NN

- Input track ip as a variable?

No use of tracking for tau identification at Tevatron, so far

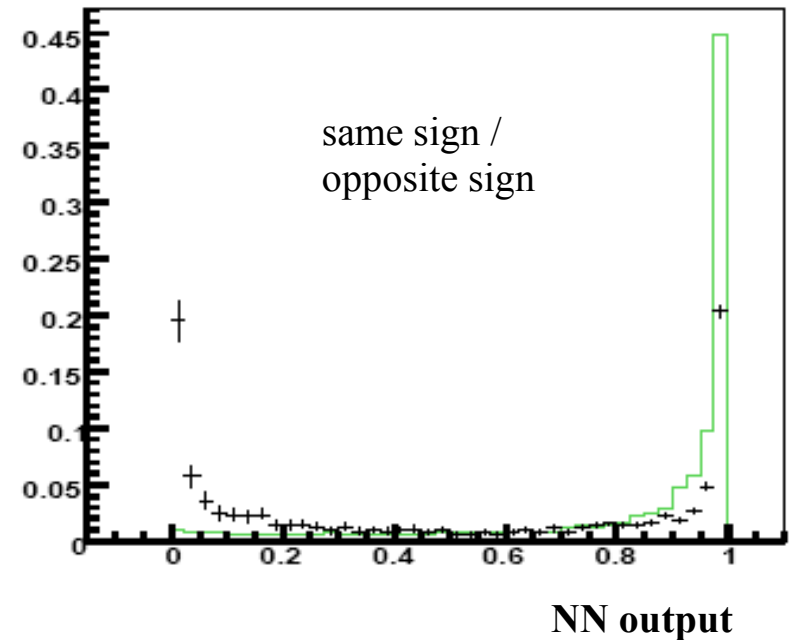
- (at D0, and I'm pretty sure at CDF)

Issues:

- Short tau lifetime
- Small tau mass
- Backgrounds from b/c jets
- Data/MC disagreements

Could be used if there were *excellent* tracking and different backgrounds

- ILC?



Conclusions / Summary

Good b/c tagging starts with good tracking!

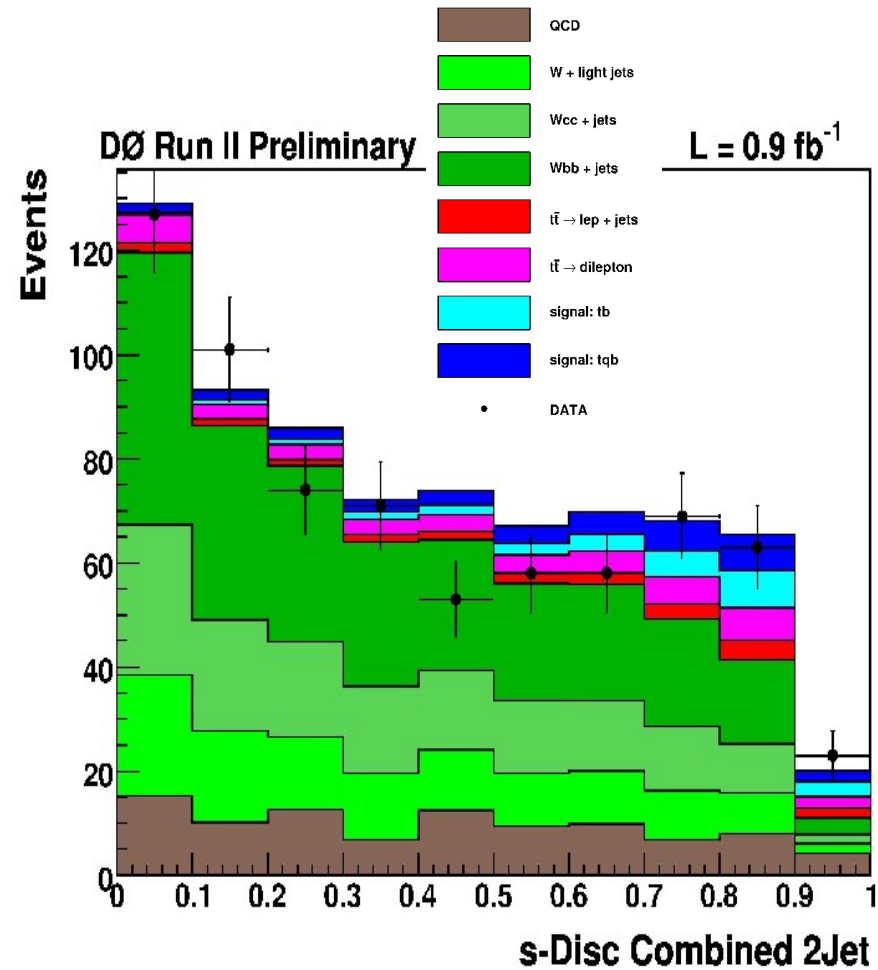
Combine jet lifetime / vertex info into NN's trained on MC to separate b/c/light jets

Measure performance in data using muon-tags / away tags / negative tags -> TRF's

Use *multiple operating points simultaneously* to take advantage of all information from tagging

Soft-lepton tagging (muons, electrons?)

tau-tagging -> for the ILC!



Evidence for single-top!

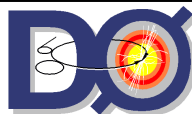
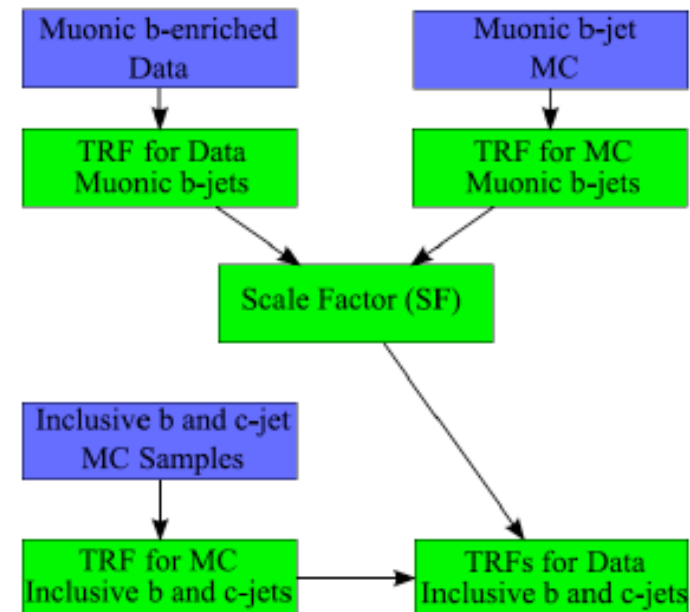
Backup

Performance estimation



- For any major software release DØ undergo the process of re-evaluating its b-tagging capability
 - **b-tagging efficiency evaluation**
 - Measurement from DATA using Muon Jets
 - Muon Jets: jets with an associated muon
 - Scale factors from Monte Carlo
 - Muon Jet b-tagging efficiency
 - c-tagging efficiency
 - Evaluation of systematic errors
 - Tag Rate Functions provided to end users for physics analysis
 - **Mistag Rate**
 - Negative tagging rate from several samples
 - QCD and EM samples
 - Correction from Monte Carlo
 - asymmetry and heavy flavor content
- **Taggability**: in order to decouple b-tagging efficiency to effects not simulated in MC (i.e. inefficiency of SMT modules) only jets associated to tracks are used for b-tagging
 - Taggability measured in DATA and used to scale MC

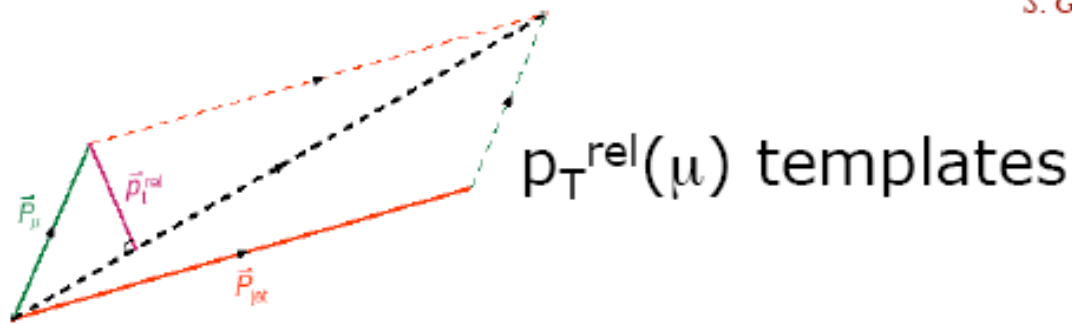
T. Scanlon Ph.D. Thesis



b-tagging efficiency from data



S. Greder Ph.D. Thesis

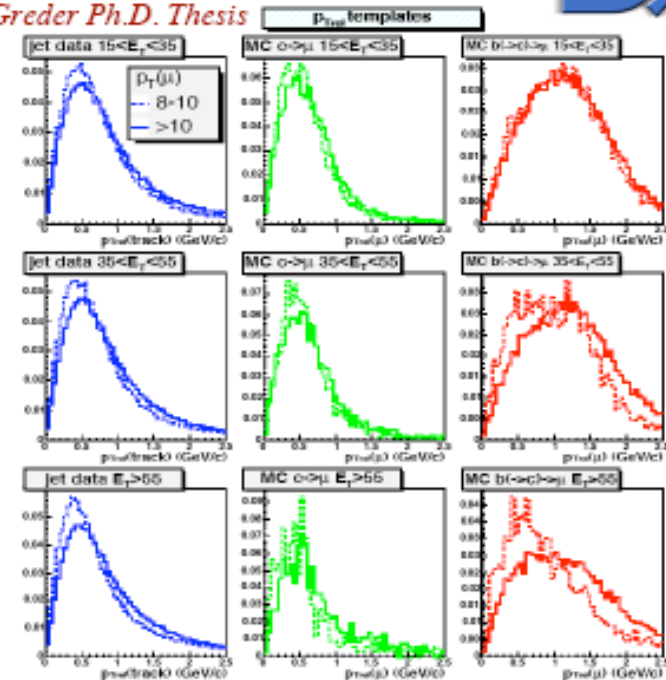


- b-tagging efficiency is estimated from Muon Jets sample (**high b-quark content**)
- b-tagging efficiency is derived from the estimation of b-quark content of the Muon Jet sample before and after applying b-tagging

- Fit of p_T^{rel} templates
- p_T^{rel} depends upon jet and muon p_T
 - Several templates used

• Currently used only for counter checks

$$\epsilon_{\text{b-tagging}} = (N_{\text{tagged jets}} \cdot f^{\text{b}}_{\text{tagged jets}}) / (N_{\text{jets}} \cdot f^{\text{b}}_{\text{jets}})$$



L.F. Ph.D. Thesis

