

Identifying b-Jets in hadronic collisions a tool for discoveries

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LIP seminar

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Caveat:

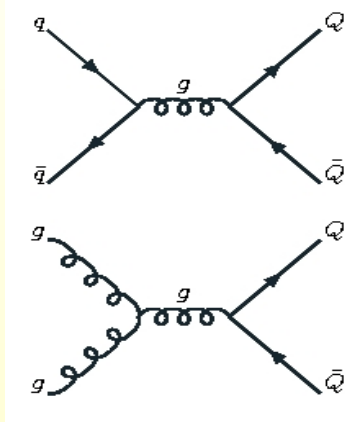
- This talk mixes information taken from my personal experience in:
 - Jet b-tagging at Tevatron and in particular in CDF
 - Some consideration on LHC Jet and b-Jet physics with ATLAS view
- I will not stress on physics results in themselves, but rather:
 - Focus on experimental strategies for b-tagging
 - Make main points connected with phenomenology
 - Have a transversal approach between experiments and physics channels.

Introduction

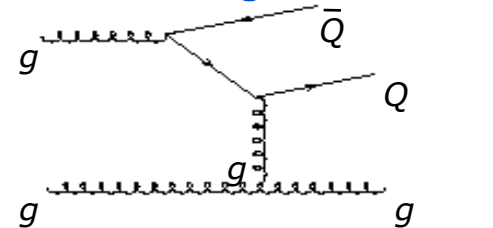
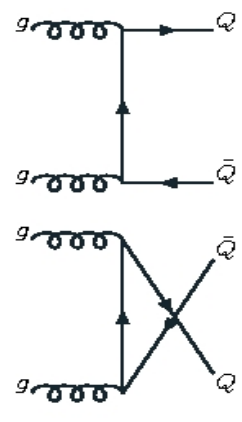
- Lots of interesting physics involves high- p_T b-quarks
 - Top production - $\text{BR}(t \rightarrow Wb) \sim 100\%$
 - Higgs searches
 - For $m_H < 135 \text{ GeV}/c^2$, $H \rightarrow b\bar{b}$ is the most common decay
 - SUSY $b\bar{b}g \rightarrow b\bar{b}A \rightarrow b\bar{b}b\bar{b}$ at high $\tan\beta$
 - Sparticle searches – sbottom, stop decays into $b+X$
- There are well established ways to b-tag and more exotic ones for future
- Outline:
 - b quarks production at collider
 - Characteristics of b-jets
 - B-tagging algorithms
 - Efficiency and fake rate measurements, the CDF experience
 - Recent algorithmic advances
 - Some highlights on the LHC
 - Conclusions

bb pairs production in hadronic collisions

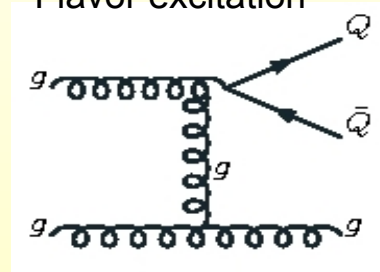
Leading Order and Next to Leading Order



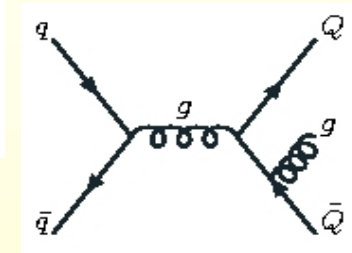
Flavor creation



Flavor excitation



Gluon splitting



and ... radiative corrections

... but: **b/c-jets** or **B/D hadrons** are the observables rather than b/c-quark

Observed

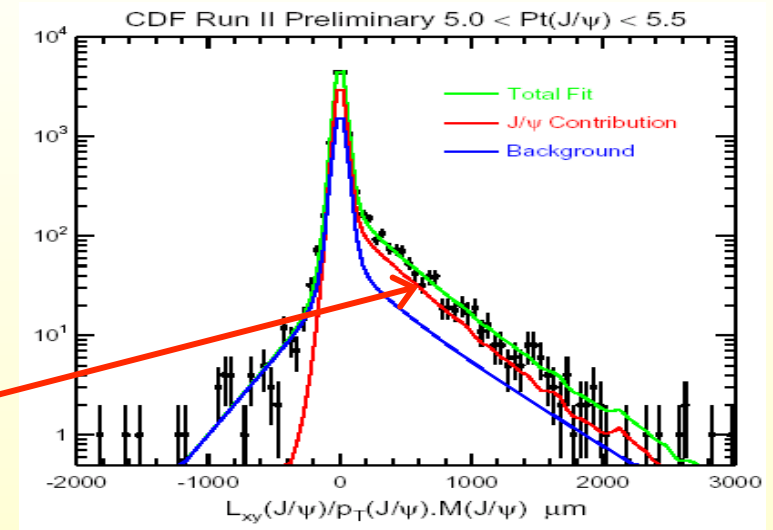
$$\frac{d\sigma(p\bar{p} \rightarrow B/DX)}{d p_T(B/D)} = \underbrace{\frac{d\sigma(q\bar{q} / gg / qg \rightarrow b/cX)}{d p_T(b/c)}}_{\text{NLO QCD}} \otimes \overset{\text{Proton structure}}{\uparrow} F^{p\bar{p}} \otimes \underset{\text{Fragmentation}}{\downarrow} D^{b/c \rightarrow B/D}$$

Factorizes into a **calculable** part and a **non-calculable** but universal piece

Inclusive b cross section - low P_T



- Run I: b cross-section $\sim 3\times$ old NLO QCD
- **Theoretical approaches**: Next-to-Leading-log resummations, non perturbative fragmentation function from LEP, new factorization schemes...
- **Experimentally**: unbinned maximum likelihood fit to the J/ψ decay time in the $R-\phi$ plane to extract the b fraction



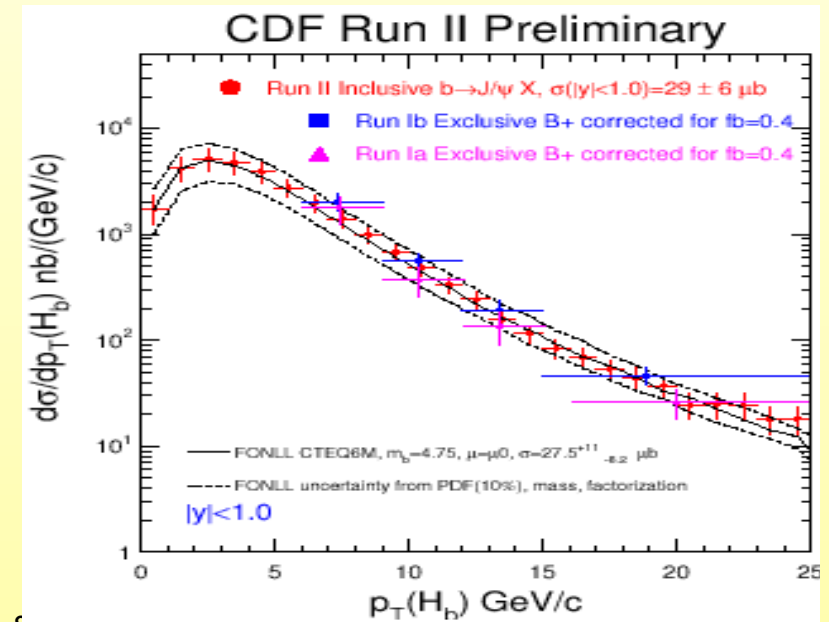
Run II: **Bottom Quark Production cross-section:**

$$\sigma(p\bar{p} \rightarrow bX)|_{|y|<1.0} = (29.4 \pm 0.6(stat) \pm 6.2(sys)) \mu b$$

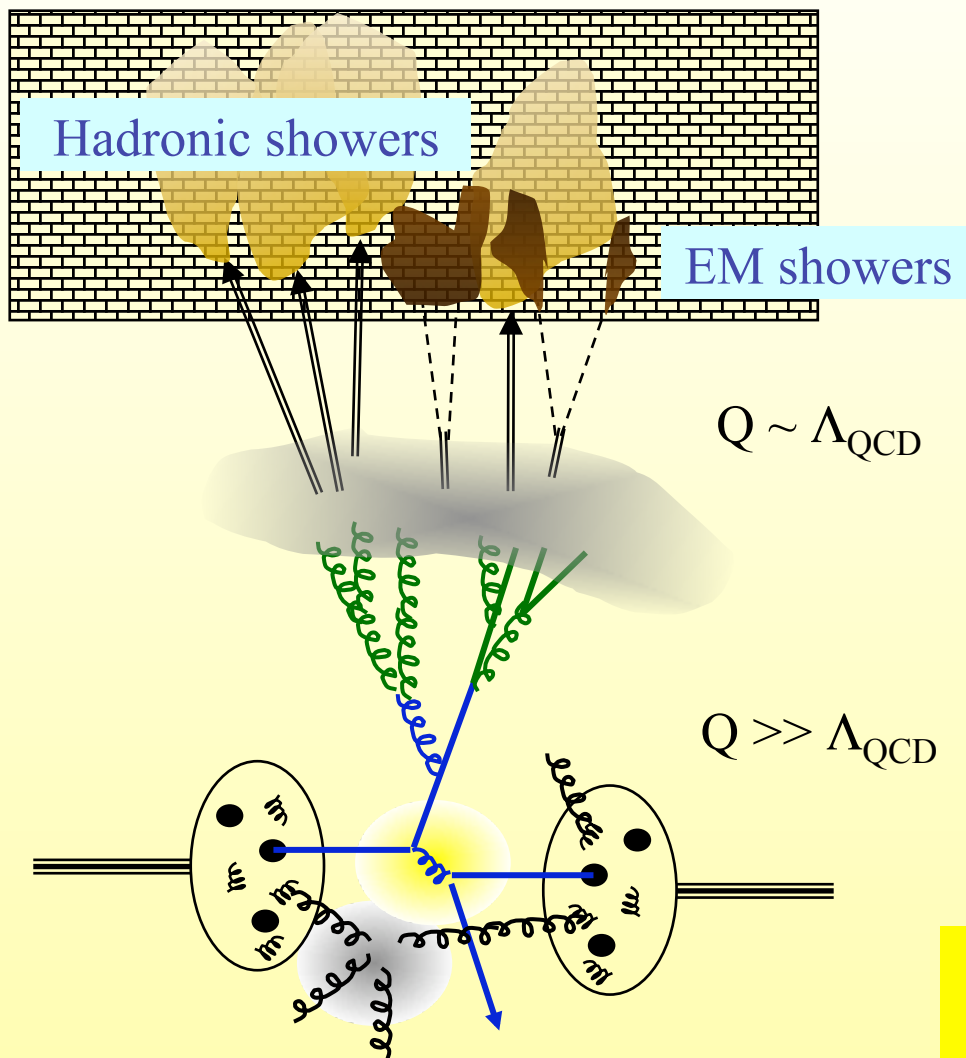
PRD 71, 032001 (2005)

Good agreement →

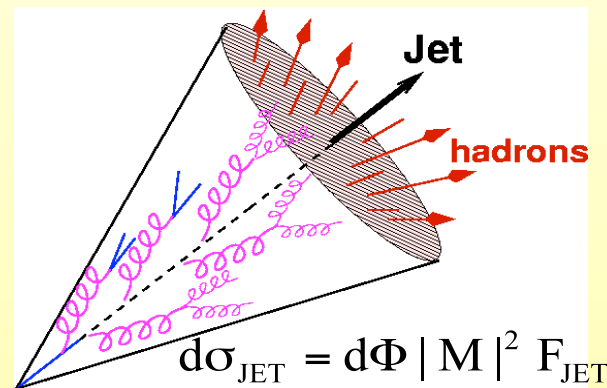
$$\text{FONLL } \sigma(p\bar{p} \rightarrow bX)|_{|y|<1.0} = (27.5^{+11}_{-8.2}) \mu b$$



Jet definitions



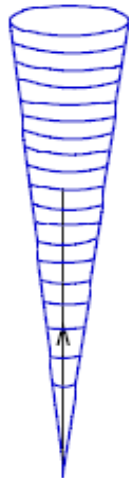
- Unfold detector to Particle level
- ↪ Correct for efficiency, resolution
- Correct theory (pQCD) for non-perturbative effects
- ↪ Underlying Event, Fragmentation



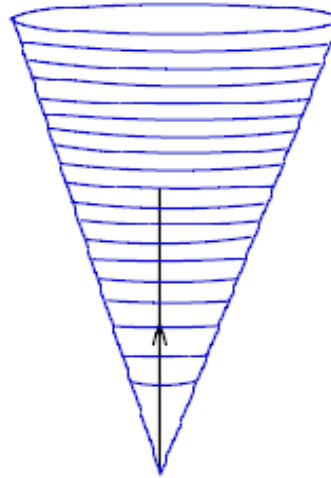
Well defined jet algorithm required

↪ **At calorimeter, hadron and parton levels**

Small jet radius

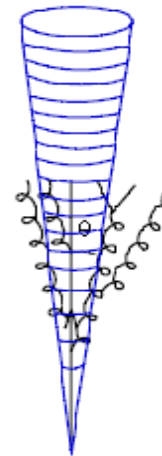


Large jet radius

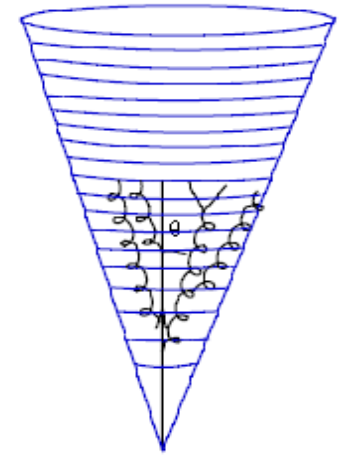


single parton @ LO: **jet radius irrelevant**

Small jet radius

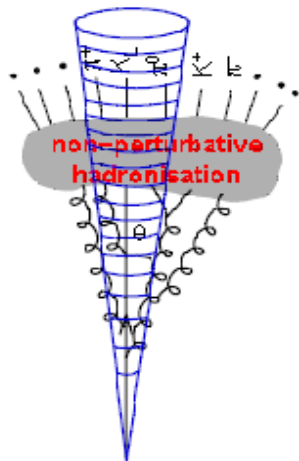


Large jet radius

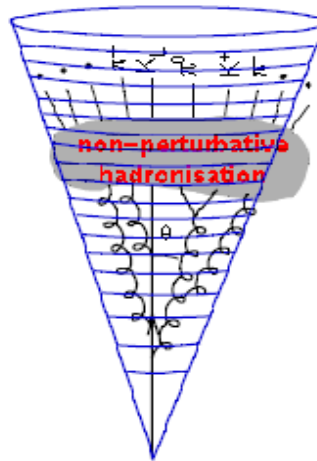


perturbative fragmentation: **large jet radius better**
(it captures more)

Small jet radius

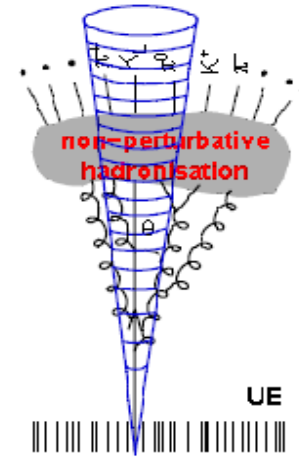


Large jet radius

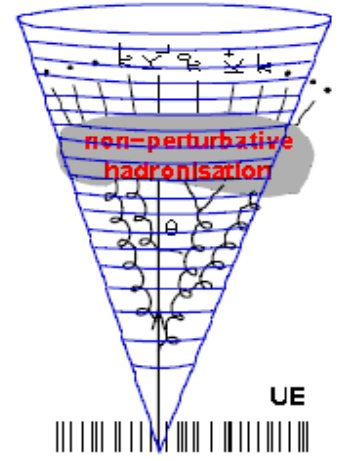


non-perturbative fragmentation: **large jet radius better**
(it captures more)

Small jet radius



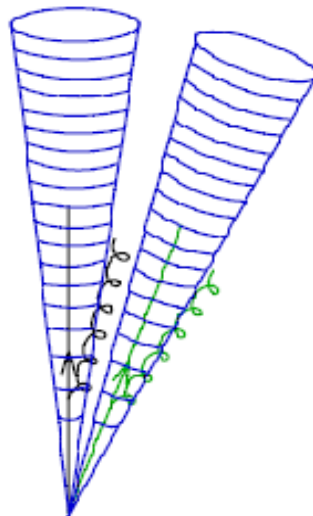
Large jet radius



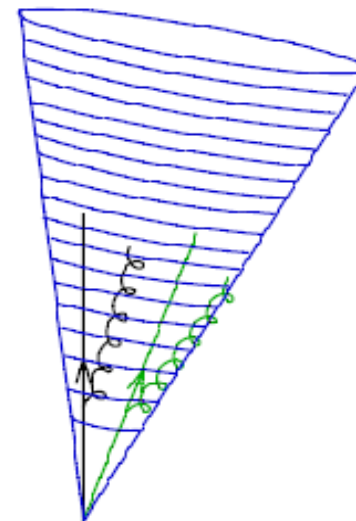
underlying ev. & pileup "noise": **small jet radius better**
(it captures less)

Jet issues

Small jet radius



Large jet radius



multi-hard-parton events: **small jet radius better**
(it resolves partons more effectively)

4-way tension in many measurements:

Prefer small R	prefer large R
resolve many jets (e.g. $t\bar{t}$)	minimize QCD radiation loss
limit UE & pileup	limit hadronisation

Jet definition choice

- ▶ A jet is not a parton: it's (sort of) what you choose it to be.
- ▶ It's easier to think in terms of partons (LO, NLO pQCD) with IR/Collinear safe jet algorithms. *And gives sense to pQCD predictions*
- ▶ \exists many cones algs. Not equivalent. Many are IR/Coll unsafe.
xC-SM \rightarrow SIScone; xC-PR \rightarrow anti- k_t
- ▶ "The best" jet definition *does not exist*
- ▶ To get the most out of jet-algs.,
 - ▶ Understand the interplay of physical scales *high $p_t \rightarrow$ larger R*
 - ▶ Try out different combinations of algorithm & R
 - ▶ Check Variations of alg. & R don't change extracted physical quantities
- ▶ Special cases (e.g. boosted $W/t/\dots$) benefit from special techniques
e.g. seq. recomb. "jet-decomposition" is a powerful tool

The Anti-Kt Algorithm

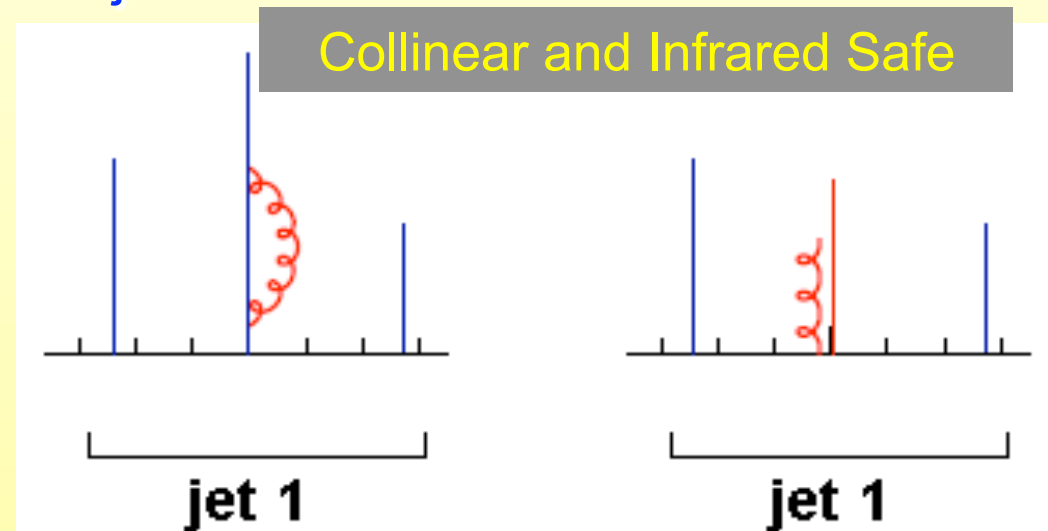
*ATLAS default now
(together with cone)*

- Infra Red and collinear Safe.
- Good Speed (Better than the SISCone)
- Behaves like an idealized cone algorithm, Conical Jets, active and passive areas are equal.
- Distance Definition:

$$d_{ij} = \min(k_{ti}^{2p}, k_{tj}^{2p}) \Delta_{ij}^2 / R^2 \quad \text{with } p = -1.$$

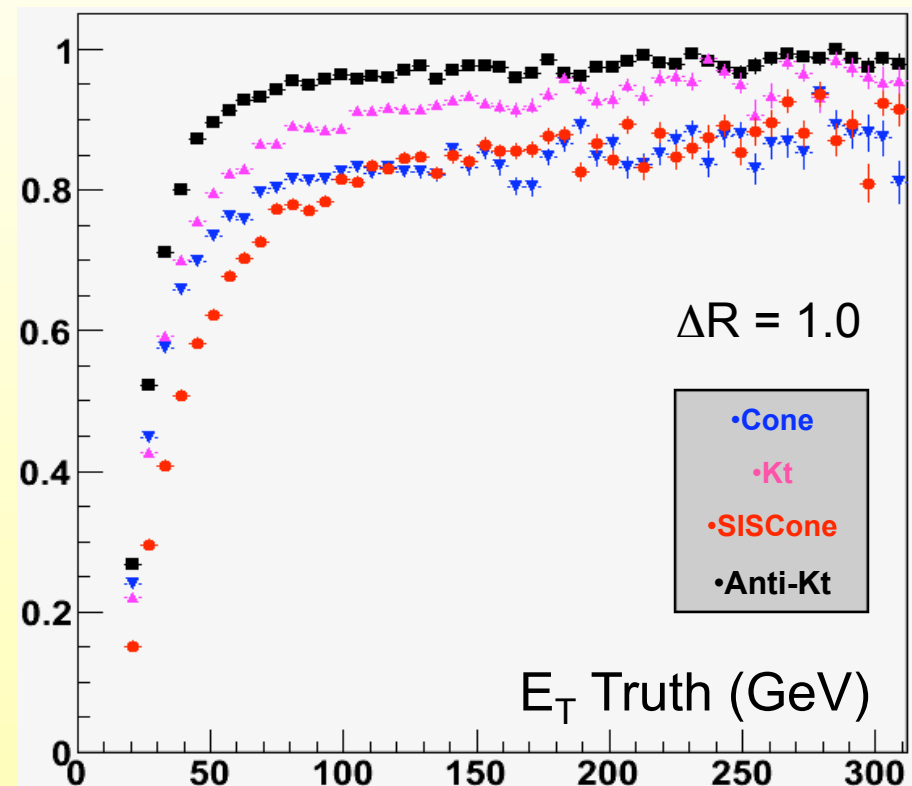
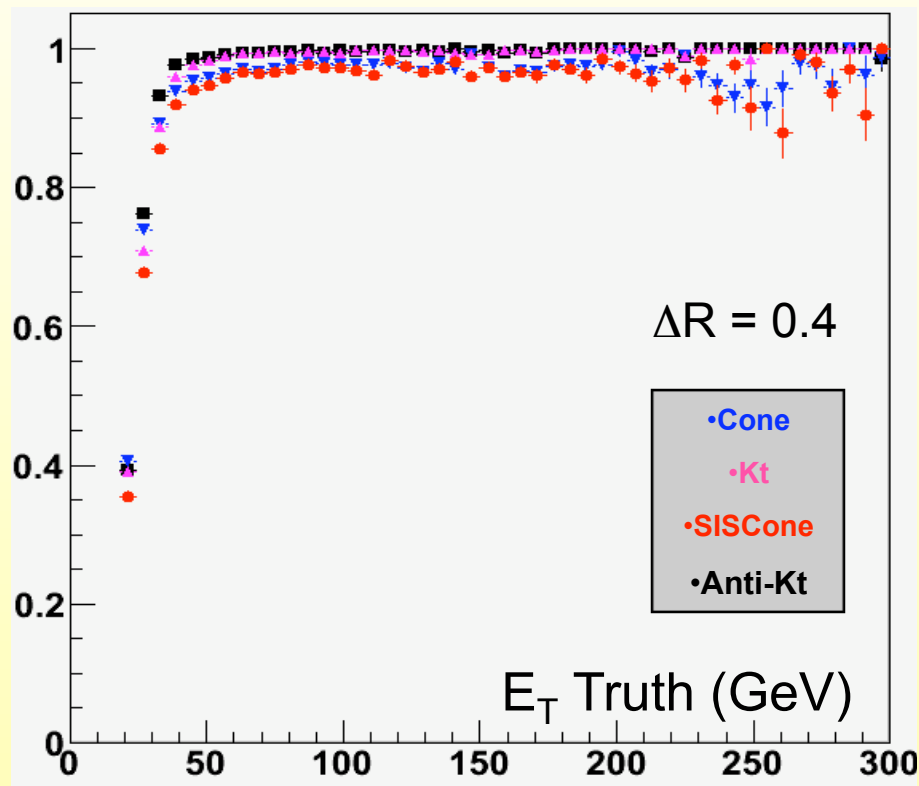
$$d_{ij} = \min \left(\frac{1}{k_{ti}^2}, \frac{1}{k_{tj}^2} \right) \frac{\Delta_{ij}^2}{R^2}, \quad d_{iB} = \frac{1}{k_{ti}^2}$$

M. Cacciari, G. Salam, hep-ph 0704.0292v2



Top Jets reconstruction efficiency

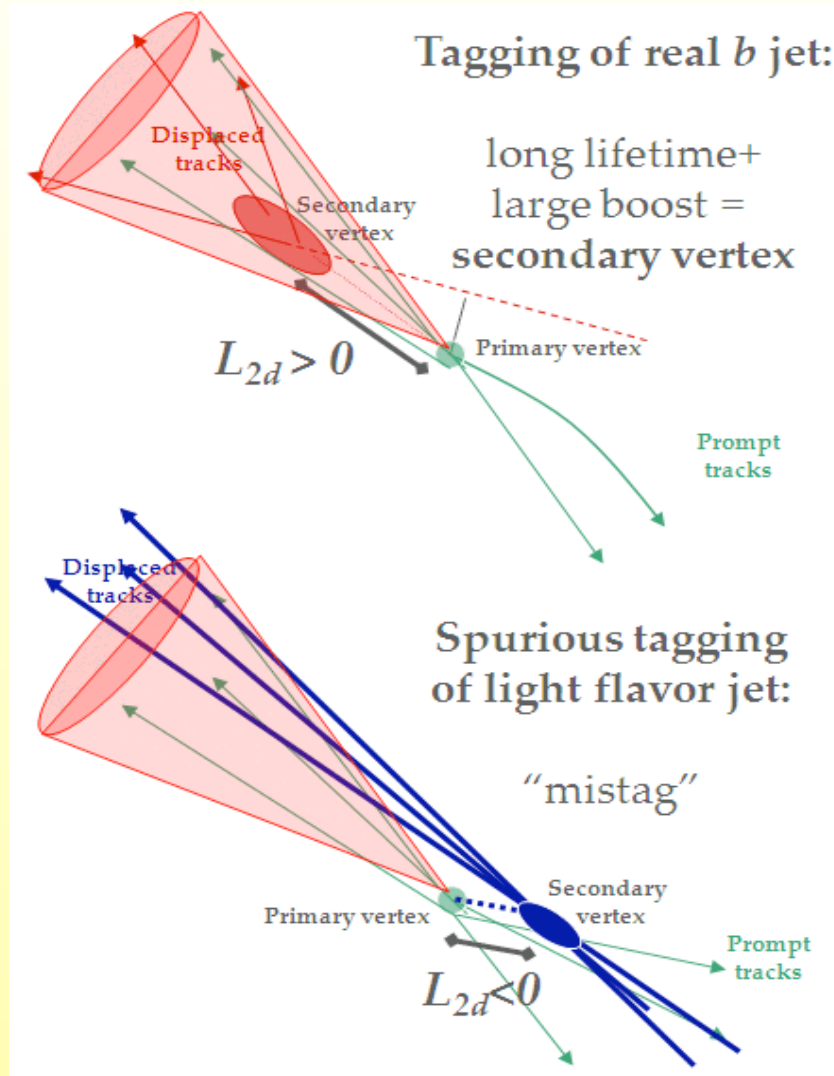
ATLAS - Jet algorithm workshop january 09 and Lisbon hadronic workshop June 09
→ contribution to Atlas default Jet algo choice



Work done at LPNHE with **HELEN** program fellow student from
Universidad de los Andes – Merida, Venezuela

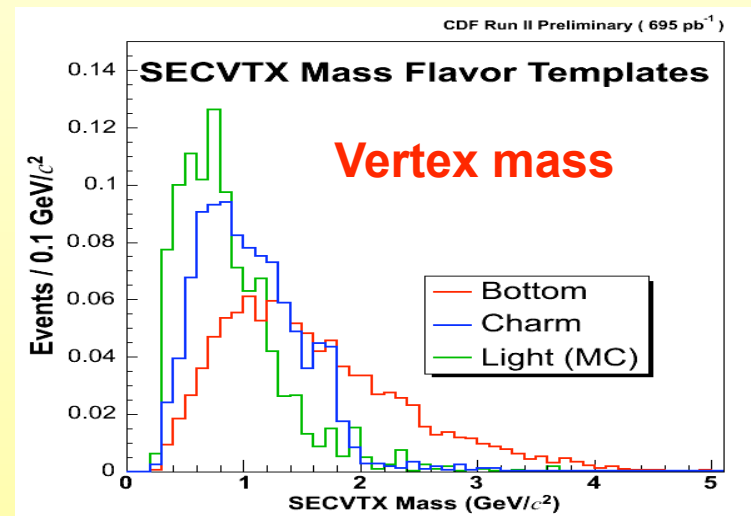
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- *To b or not to b ?*

High P_T Jets and b tagging



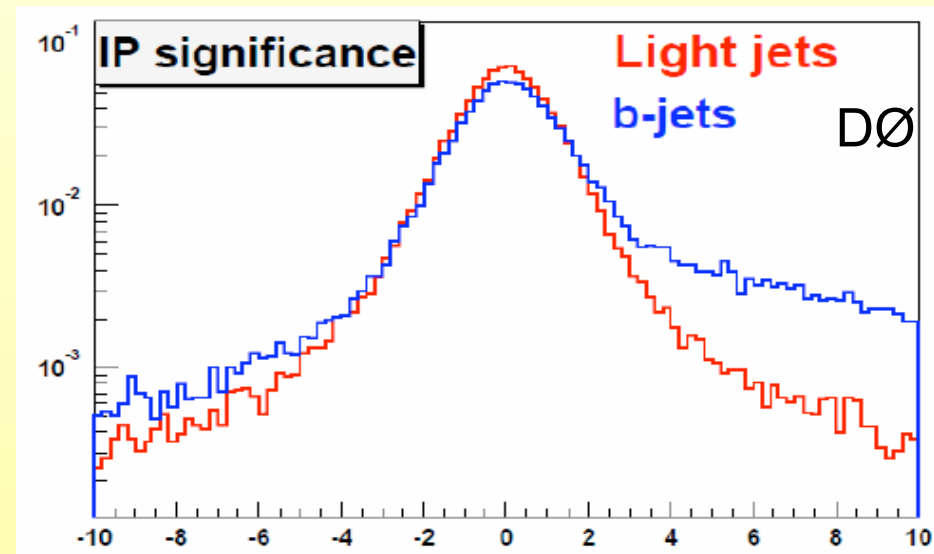
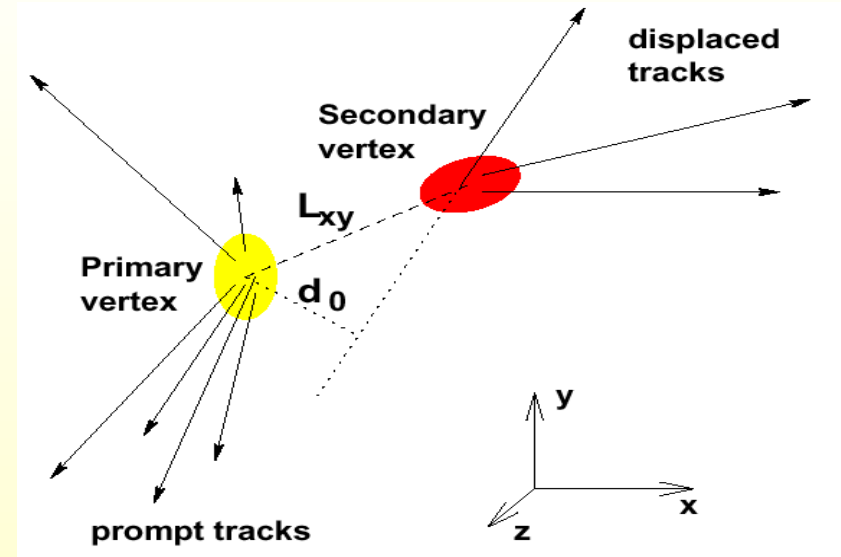
Secondary Vertex b-tag Algorithm (CDF SecVtx)

- Select tracks within the Jet Cone
- Reconstruct 2 or 3 tracks vertex
- Cut on vertex displacement significance to reduce light Jets rate
- **Secondary vertex mass** has some separating power between light, charm and b Flavors, as like as other variables



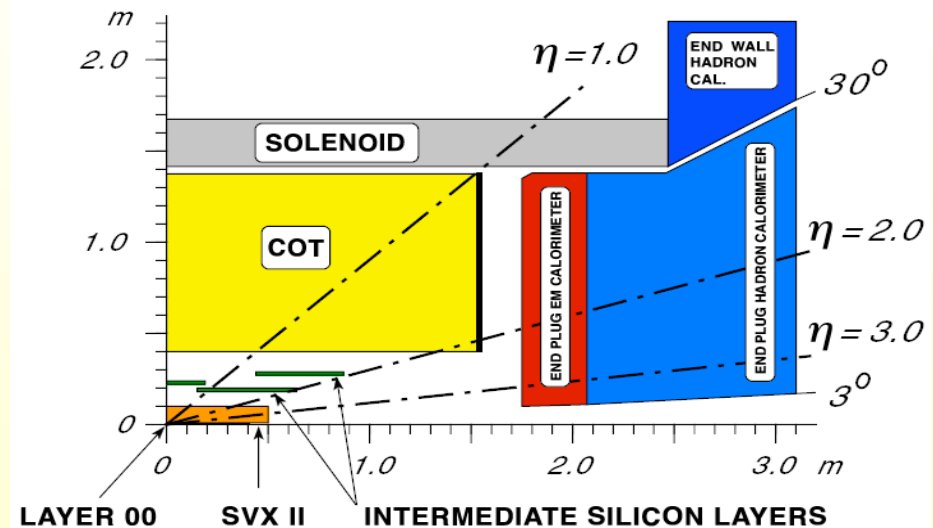
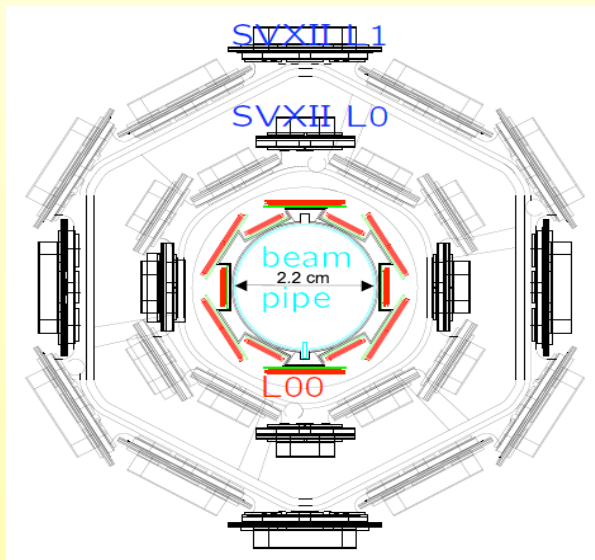
Signatures of B-Jets

- High b-hadron mass
 - $5.3 \text{ GeV}/c^2$
- Relatively long lifetime
 - $1.5 \text{ ps} \rightarrow c\tau = 450 \text{ }\mu\text{m}$
- Hard fragmentation
 - b-hadron retains $\sim 70\%$ of b-quark momentum
- Put 'em all together and you get
 - High- p_T tracks
 - Large impact parameters
 - Secondary vertices (few mm)
- Lepton production
 - $\sim 10\% \text{ } b \rightarrow \ell \nu$
 - Also $\sim 10\% \text{ } c/\bar{c} \rightarrow \ell \nu$
 - High- p_T tracks
 - High- p_T relative to b-jet



The CDF Detector

- COT: 96-layer wire drift chamber
 - 4 axial, 4 stereo superlayers
 - 1.4 T magnetic field
- “Central” lepton ID out to $|\eta| < 1.1$
- Silicon tracker
 - 95 cm long Layer 00
 - Radiation hard
 - 3 SVX-II barrels, each 29 cm long



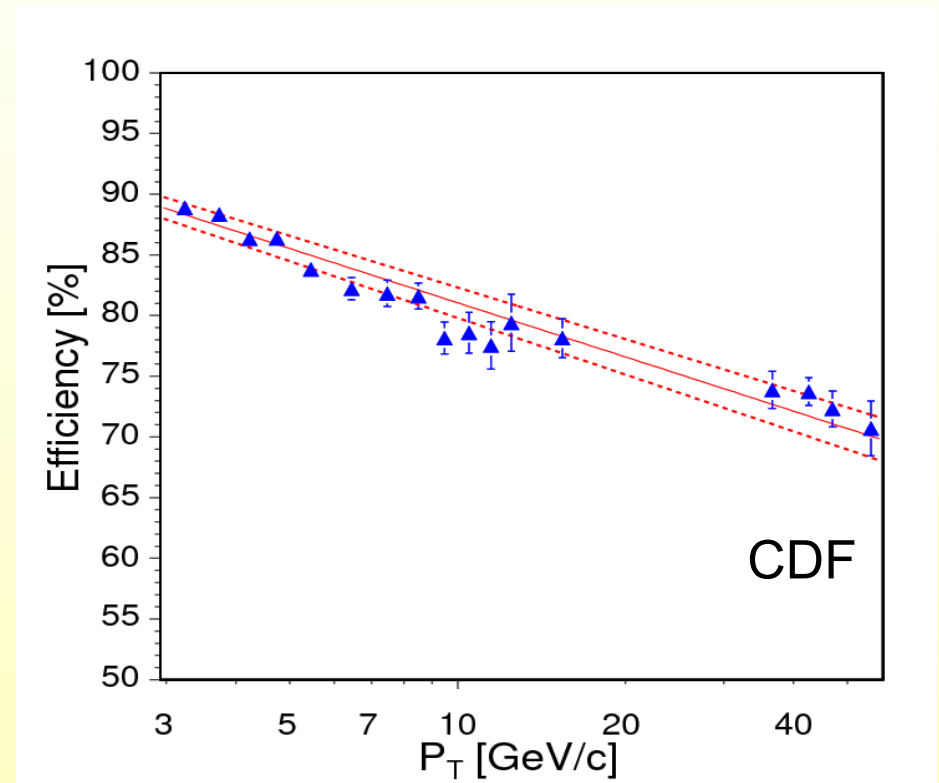
- L00 – singled-sided, $r = 1.2$ cm
- SVX-II – 5 double-sided layers
 - L0: $r\phi - rz$, $r=2.5$ cm
 - L1: $r\phi - rz$, $r=4.5$ cm
 - L2: $r\phi - 1.2^\circ$ stereo, $r=6.5$ cm
 - L3: $r\phi - rz$, $r=8.5$ cm
 - L4: $r\phi - 1.2^\circ$ stereo, $r=10.5$ cm
- Total sensor area 6 m²
- 720k electronics channels

Detector Issues

- Reference for impact parameter calculations
 - Tevatron beam profile is $\sim 30 \mu\text{m}$ in xy
 - Can improve by computing event-by-event primary vertex
 - Fit all tracks to a common point – discard outliers – iterate
 - Final resolution is $10\text{-}30 \mu\text{m}$ depending on event topology
- Poorly-reconstructed tracks are a killer – quality cuts are critical
 - Both experiments use silicon hit multiplicity and fit χ^2
 - Remove K_S/Λ decay products
 - Raising p_T cuts helps control fake tag rate
- Inactive detector regions and data/simulation mismatch
 - CDF: model inactive silicon ladders in the simulation run-by-run
 - DØ: everything relative to “taggable” jets (couple of silicon tracks) – get taggability rate from data ($\sim 80\%$)
 - Both experiments apply a scale factor to simulation efficiency to match what’s seen in the data ($0.7 - 0.9$, depending on tagger)

Soft Lepton Tagging

- Tag b-tags by identifying a lepton from b or c decay
- CDF & DØ both do it for muons
- ID requirements somewhat different than high- p_T case
 - Can't use calorimeter energy
 - Track-stub angular matching is more effective
- Typical ID efficiency 80-90%
 - Per-jet efficiency ~10%
- Mis-ID rates ~0.5% per track
- Electrons are more difficult
 - Had it in Run I
 - Under development at CDF

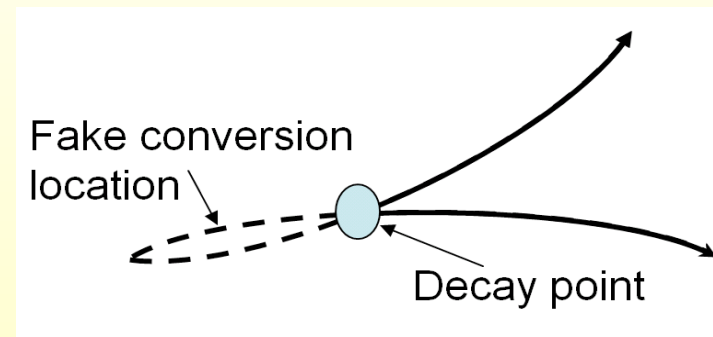
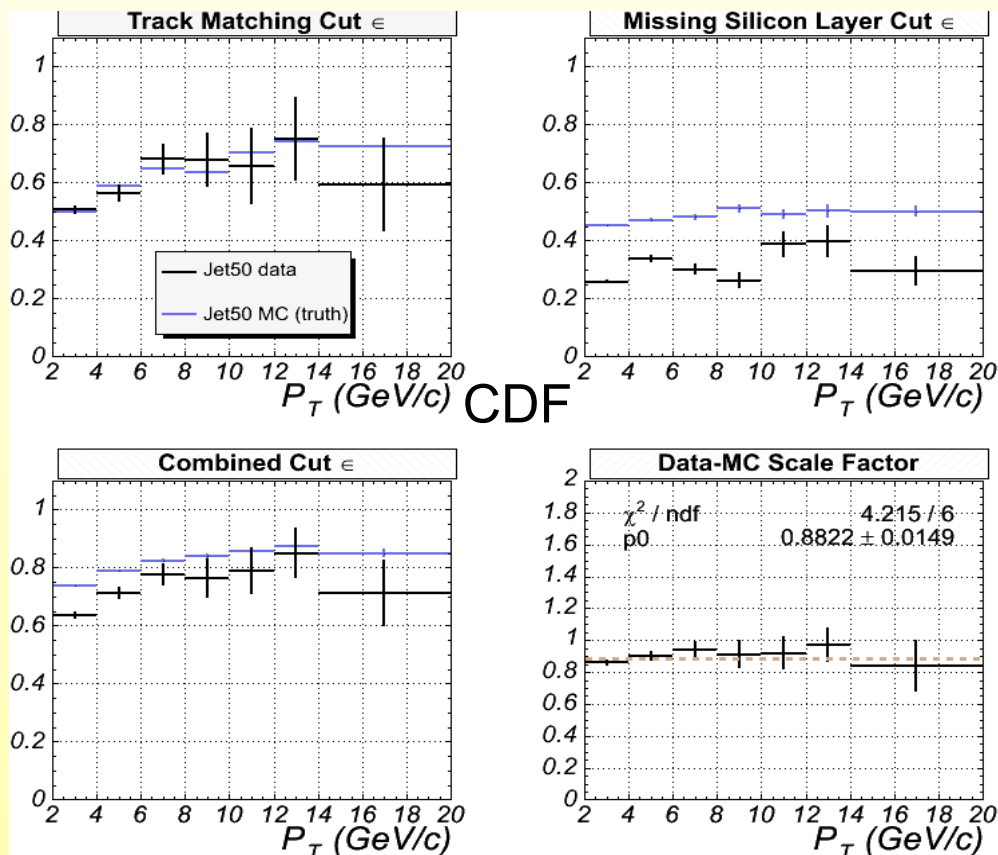


Soft lepton taggers use different information than lifetime taggers

Can use overlap rate for calibration

Soft Electron Tagger

- Additional information looking for "soft" ($p_T > 2\text{GeV}$) HF electrons within Jets
- Conversions electrons are main background
- Complementary soft electron tagged samples for tagger performance studies

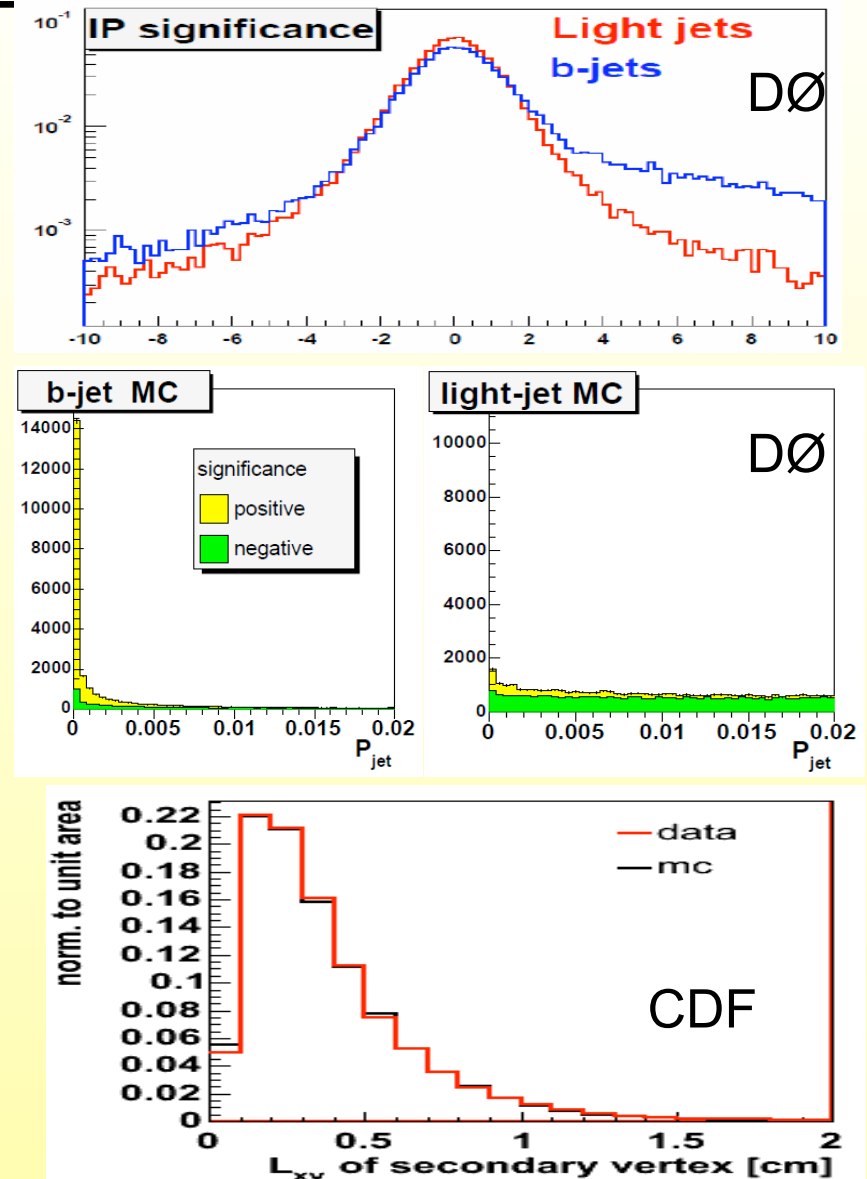


Need model for conversions
to identify them:
use partner legs

-In development also in ATLAS
exploiting fine ecal segmentation
and excellent tracking system.

Impact Parameter Taggers

- Count displaced tracks ($D\emptyset$)
 - three 2σ or two 3σ
 - sign IP's against jet direction
- Jet probability (CDF & $D\emptyset$)
 - Joint probability for all tracks to come from primary vertex
 - Track resolution derived from negative IP tracks
 - Use only positive IP tracks in probability calculation
- Displaced vertex (CDF & $D\emptyset$)
 - Fit displaced tracks (above p_T cuts) to a common vertex
 - Prune tracks and cut on fit χ^2
 - Cut on L_{xy} significance
- All algorithms can be tuned by adjusting cuts – 2-6 operating points

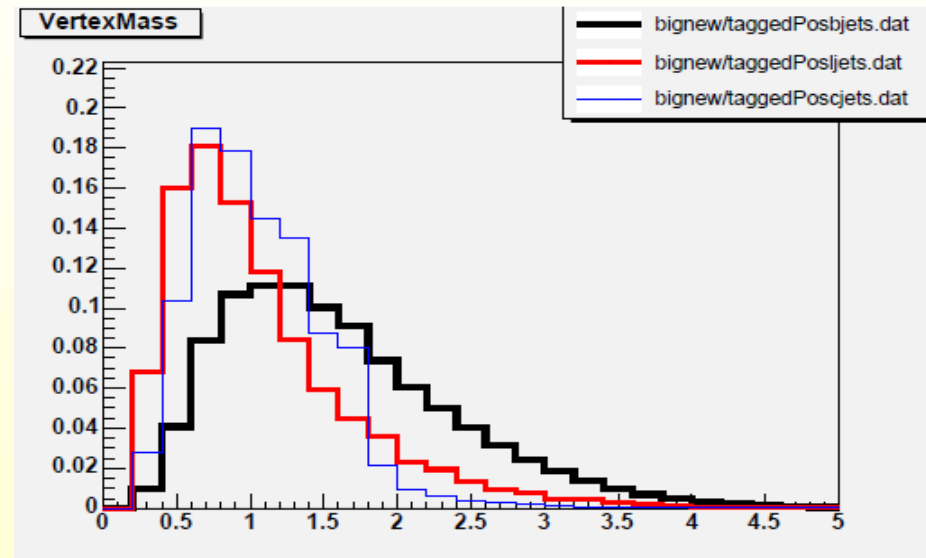


Multivariate Algorithms

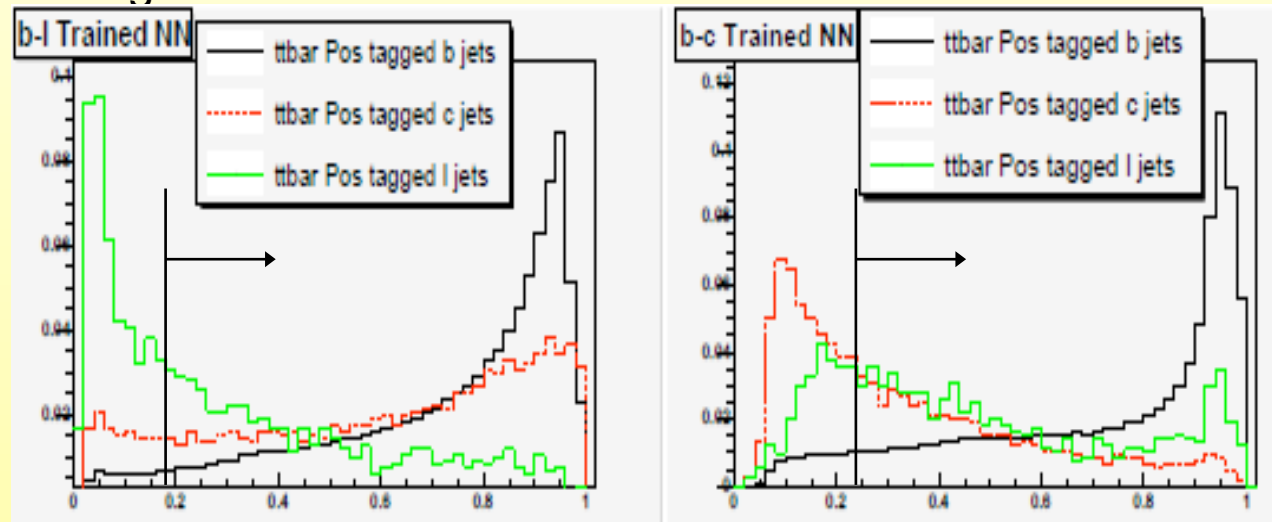
- Taggers are correlated, but not 100%
 - Can gain by combining them
 - CDF has a simple “combined tagger” – logical OR of displaced vertex and jet probability taggers
 - Gain 15-25% efficiency, at cost of factor two mistag rate
- Use more information than just the tagger outputs
 - Displaced vertex tagger gives you more than just yes/no
 - Many vertex properties discriminate signal/background
- Both experiments have new multivariate taggers
 - CDF: start with displaced vertex tag, try to reject charm/light while preserving b-tags
 - DØ: no displaced vertex prerequisite – can optimize for better purity or enhanced efficiency

Secondary Vertex + Multivariate Tagger (CDF)

- Two 16-variable neural networks
 - Vertex mass, L_{xy} , χ^2 , p_T
 - High-IP track p_T , multiplicity
 - Jet probability
 - and so much more!
- Train one to separate b-vs-light, the other b-vs-c
- Choose cuts to preserve 90% of b
 - Reject 45% of c, 65% of light



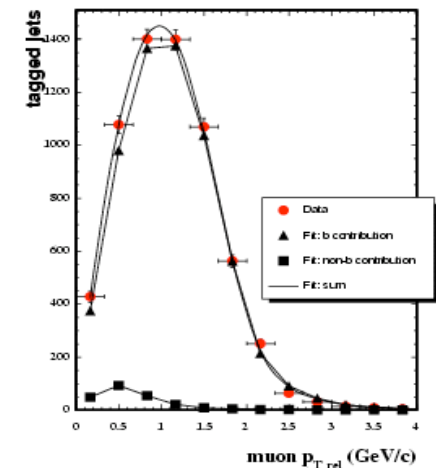
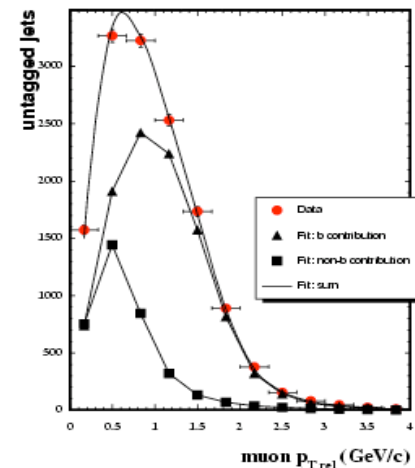
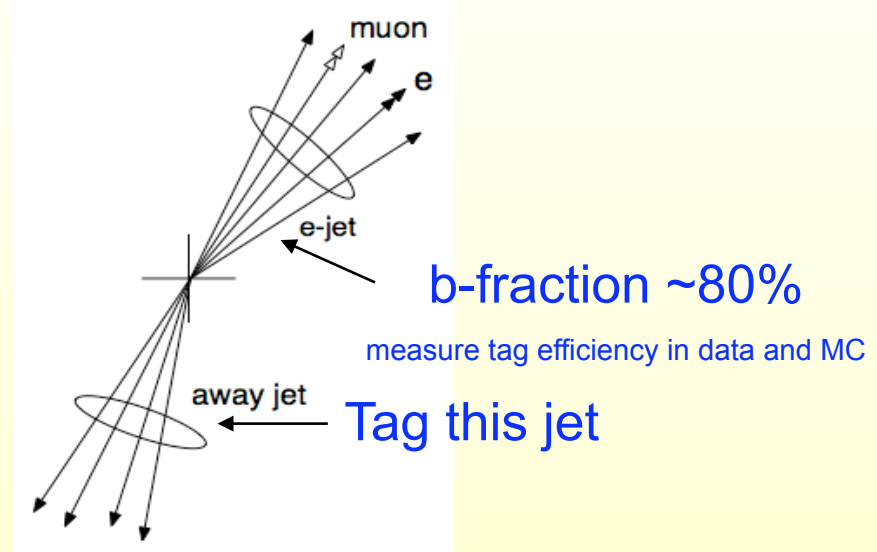
- Already in use
- Top cross section
- WH search
- Similar, dedicated algorithm for single-top search



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- *Evaluating the performances of b-tagging
the Tevatron experience on impact parameter tags*

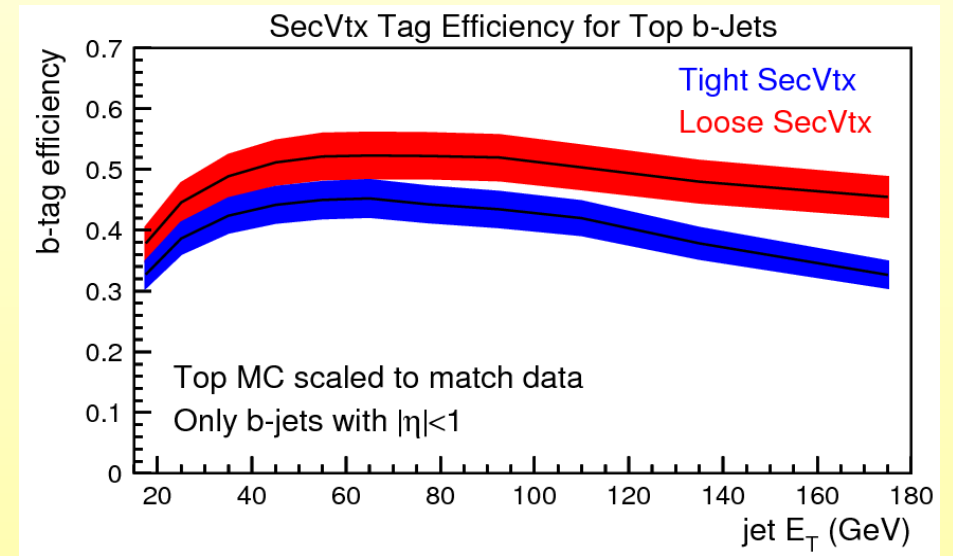
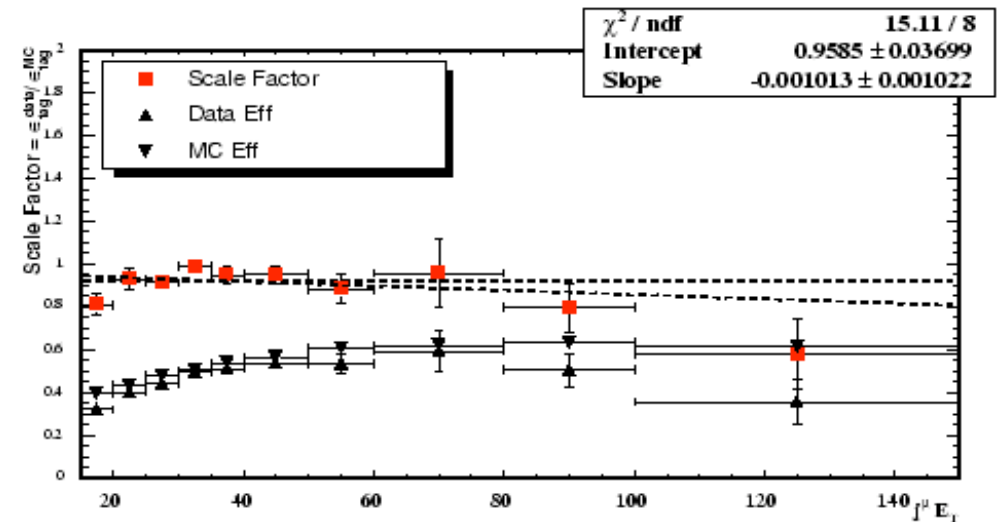
Efficiency Measurement (CDF)

- Based on 8 GeV/c electron and muon data samples
- Tag away jet to enhance b-fraction
- Generate matching MC samples
- Method A: muon p_{Trel} fits
 - Fit muon p_{Trel} against jet in tagged and untagged jet
 - Extract numbers of b-jets in each and compute efficiency
 - Systematics $\sim 3\%$, mostly from modeling of p_{Trel} templates
- Method B: electron double-tags
 - Infer non-b component from electron jet single-tag rate and conversion sample
 - Systematics $\sim 5\%$, mostly from mistag subtraction



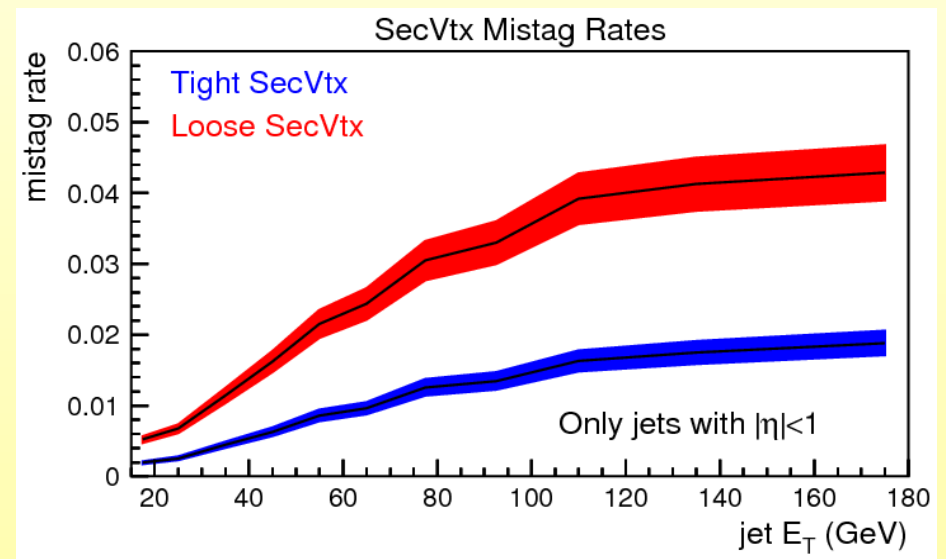
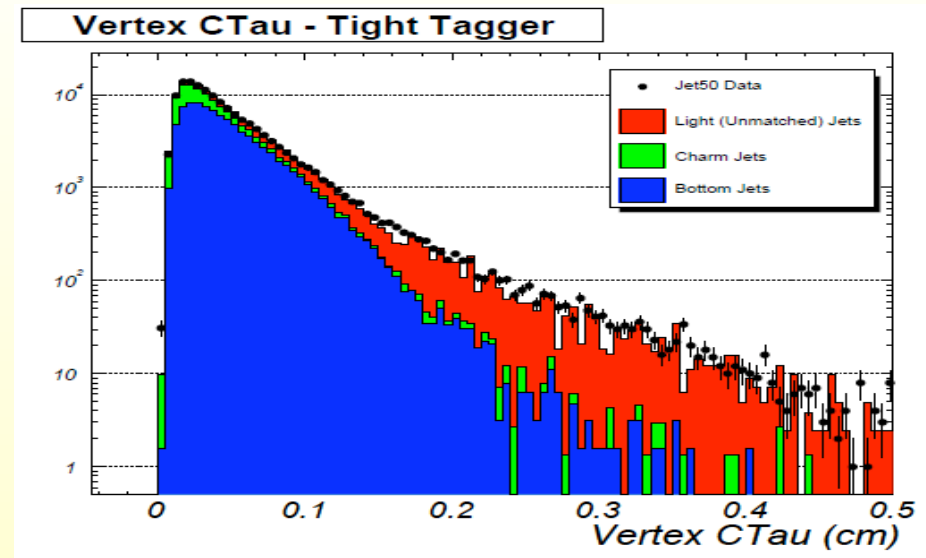
Efficiency Measurement (CDF)

- Systematics quoted earlier for each method were only the “internal” ones
- Additional systematics related to extrapolation to physics samples (in particular, b-jets from top)
- Jet E_T dependence
 - Convolute binned scale factor with E_T spectrum from top
 - 7% uncertainty
- Similar procedure for jet η
 - 1% uncertainty
- 2% uncertainty assigned for differences between semileptonic/generic B-decays
- Total uncertainty on data/MC scaled factor : 7.3%



Fake Tag Rates

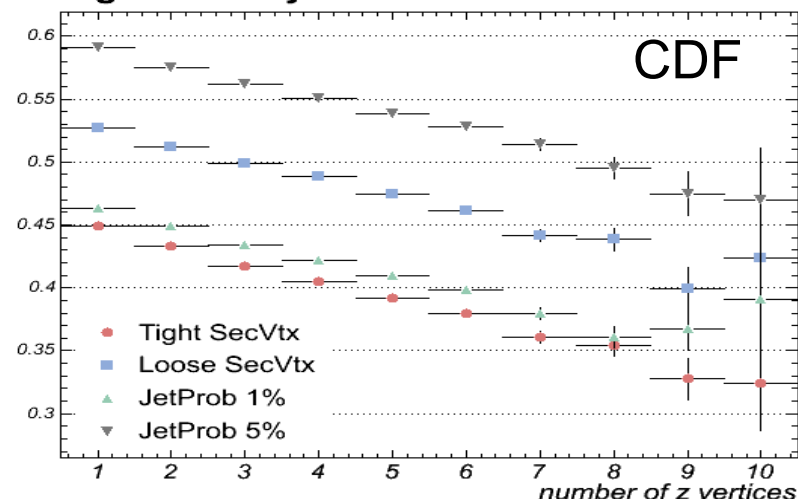
- Fake tags (mistags) mostly from misreconstructed tracks
 - +/- symmetry
- Estimate fake rates by
 - Using tracks with negative signed impact parameters
 - Using displaced vertices behind the primary w.r.t. the jet direction
- Both experiments use parametrizations of “negative” tag rates (in E_T , η , etc) to predict mistag rate in data samples
- Complications
 - Negative tags from heavy flavor (~10-15%)
 - Mistag rate not exactly symmetric – stray K_S/Λ and interactions with detector material
 - Fit POS-NEG excess in pseudo- $c\tau$ to estimate (~40%)
 - Net effect ~30% enhancement over negative tag rate



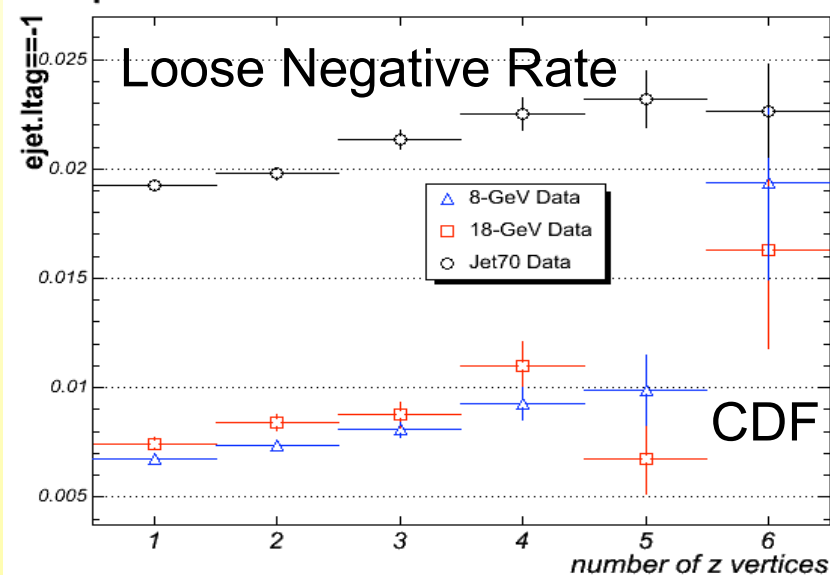
B-Tagging at High Luminosity

- Starting to study effects of high luminosity on b-tagging (results are extremely preliminary)
- $N_{\text{vtx}} \propto \text{Luminosity}$
 - At $3\text{E}32 \text{ cm}^{-2}\text{s}^{-1}$, $\langle N_{\text{vtx}} \rangle = 4$
- MC study of top events with extra minimum-bias overlaid indicates that efficiency decreases
 - High tracker occupancy
 - Hits merge together
 - Pattern recognition harder
- Negative tag rates in data rise
 - Silicon hit merging?
 - Another interaction nearby?

b-Tag Efficiency



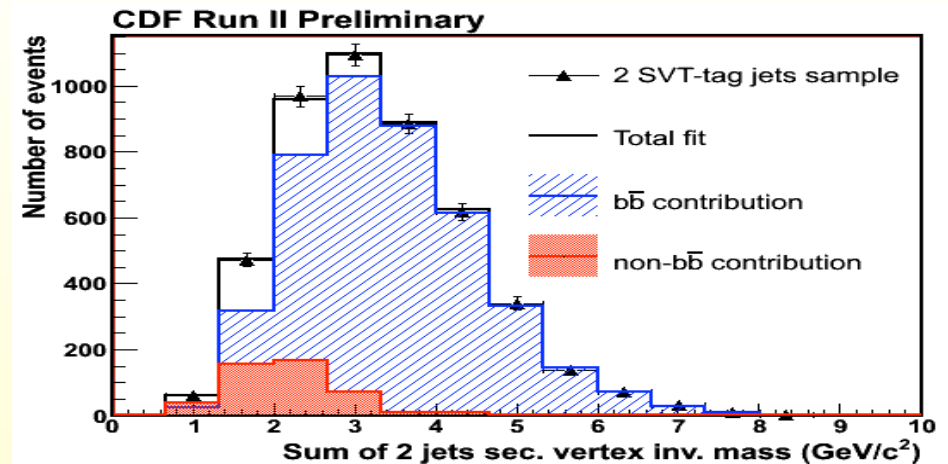
Dependence on Number of Z Vertices



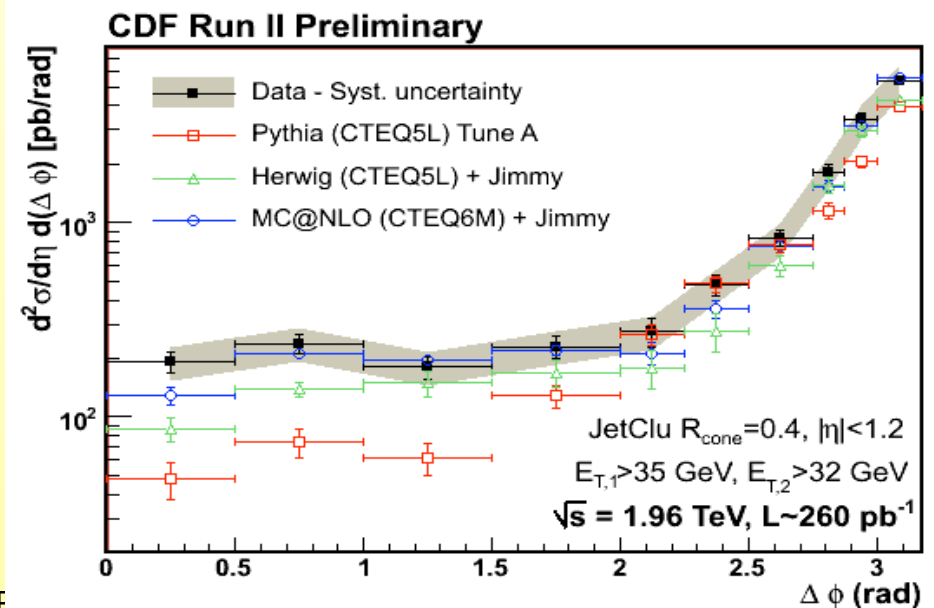
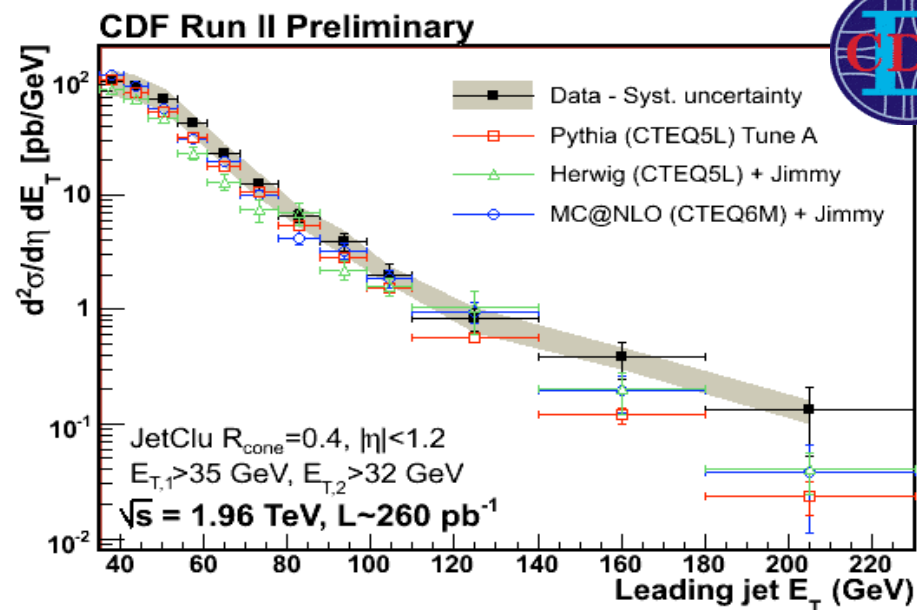
-
- ... *or directly triggering on b -Jets*

Inclusive bb di-jets production

- Specific Trigger based on L2 SVT (Secondary Vertex Trigger) used.
- Sensitive to the different production mechanisms
 - Flavor creation at high $\Delta\phi$
 - Flavor excitation or gluon splitting at low $\Delta\phi$



Purity $\sim 85\%$: extracted from data using shape of secondary vertex mass



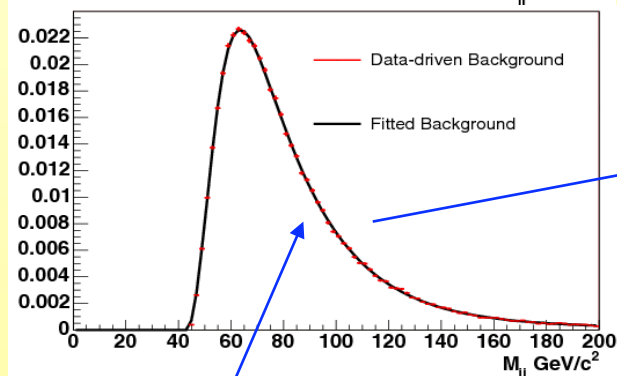
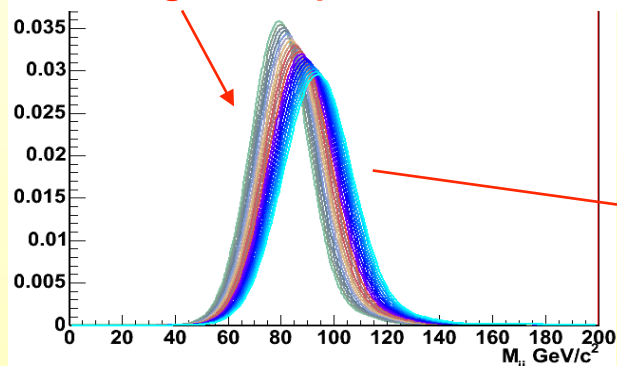
b-Jets Energy Scale from $Z \rightarrow bb$

- Generic Jet Energy Calibration of the Detector needs specific correction for b-Jets
- Reduction of uncertainty in b-JES important for Top mass measurement
- Test algorithms to improve b-jet energy resolution is crucial for low mass Higgs search.
- Use of L2 SVT based b-tag di-jet trigger \rightarrow Extract a signal, measure data/MC b-JES



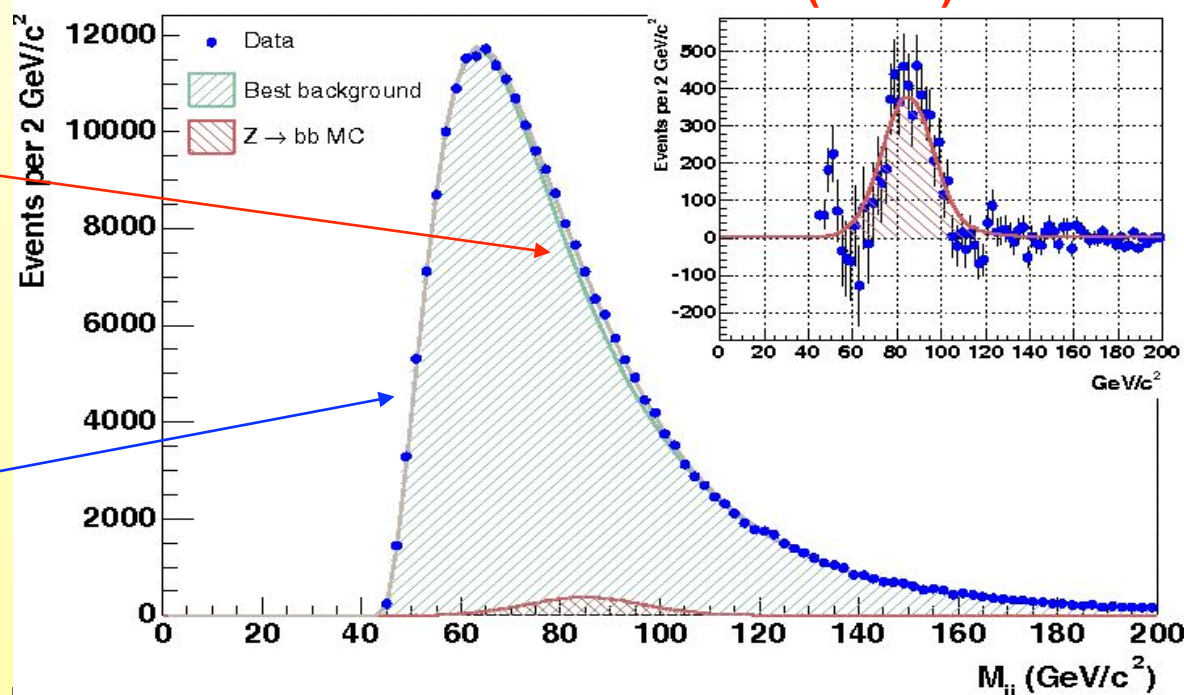
Result on $600 \text{ pb}^{-1} \rightarrow$ b-jet Data/MC Energy Scale Factor = 0.974 ± 0.020

$Z \rightarrow bb$ signal templates for $0.9 < SF < 1.09$



Background template

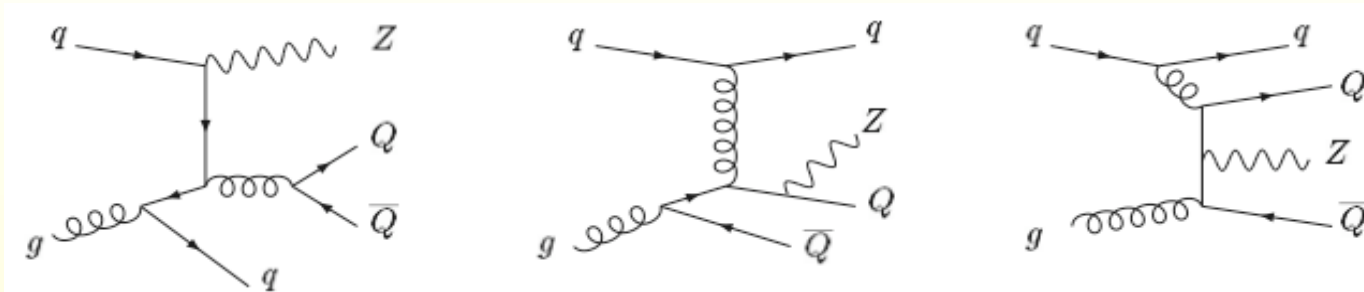
CDF Run II Preliminary $L=584 \text{ pb}^{-1}$



$N(Z \rightarrow bb) \sim 5600$

-
- *from inclusive b -Jets production...*
 - *... to gauge boson + HF associated production*
 - *Main background to Top, Higgs and NP searches*
 - *→ need for precise knowledge of b -tag performances*

W / Z / γ bosons + HF Jets production



The study of $W / Z / \gamma$ + Jets production is very relevant:

- to QCD:

- High q^2 ($\sim M_{\text{boson}}$) interactions, perturbative theory
- Less leading diagrams (wrt pure Jets production)
- Test Monte Carlo generators (ALPGEN...)
- Probe for protons PDF (in particular W/Z + H.F.)

- for many high p_T analysis where W/Z +HF is a main background:

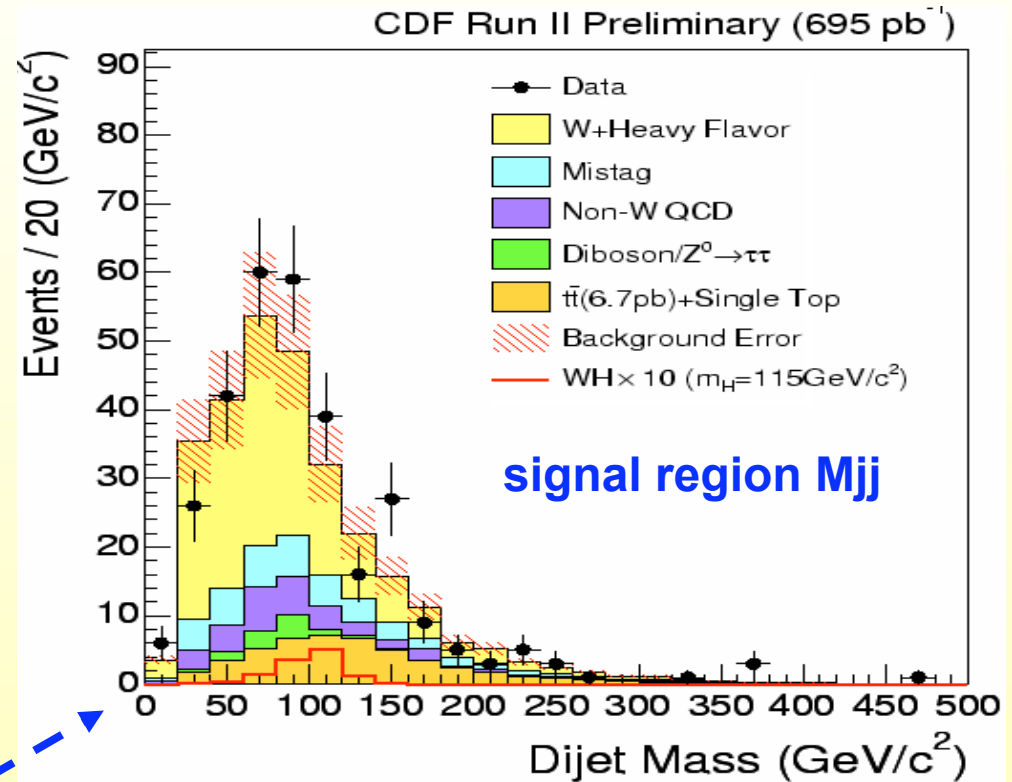
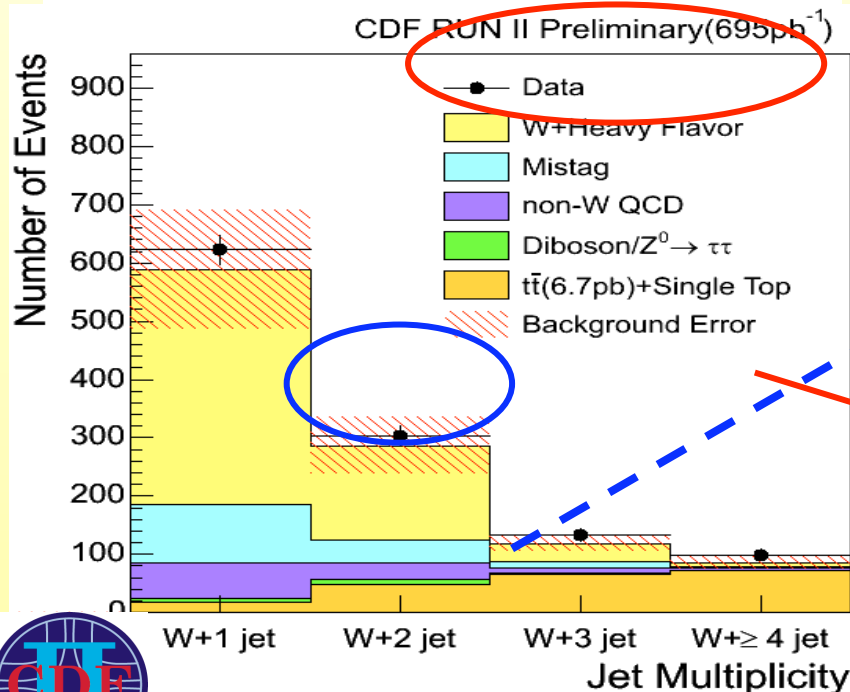
- ✓ Top quark cross section and mass
- ✓ Single Top cross section
- ✓ Search for low mass Higgs boson
- ✓ Several SUSY searches

... and plays crucial role also at LHC !

W+HF backgrounds in HIGGS WH \rightarrow $l\nu$ bb

= 1 jet tagged jet sample

Jet Multiplicity	w/o NN tag	w/ NN tag
Before b -tagging	10647	10647
Mistag	119.4 \pm 19.4	41.8 \pm 9.0
$Wb\bar{b}$	130.4 \pm 44.6	120.2 \pm 41.1
$Wc\bar{c}$	48.9 \pm 16.7	33.7 \pm 11.5
$W\tau$	47.4 \pm 12.4	25.0 \pm 6.5
$t\bar{t}$ (6.7pb)	43.1 \pm 7.3	37.8 \pm 6.4
Single Top	22.7 \pm 2.4	20.1 \pm 2.1
Diboson/ $Z^0 \rightarrow \tau\tau$	17.0 \pm 2.5	10.6 \pm 1.7
non- W QCD	44.4 \pm 7.7	29.5 \pm 5.1
Total Background	473.4 \pm 66.9	318.8 \pm 54.7
Observed(≥ 1 tag)	514	332

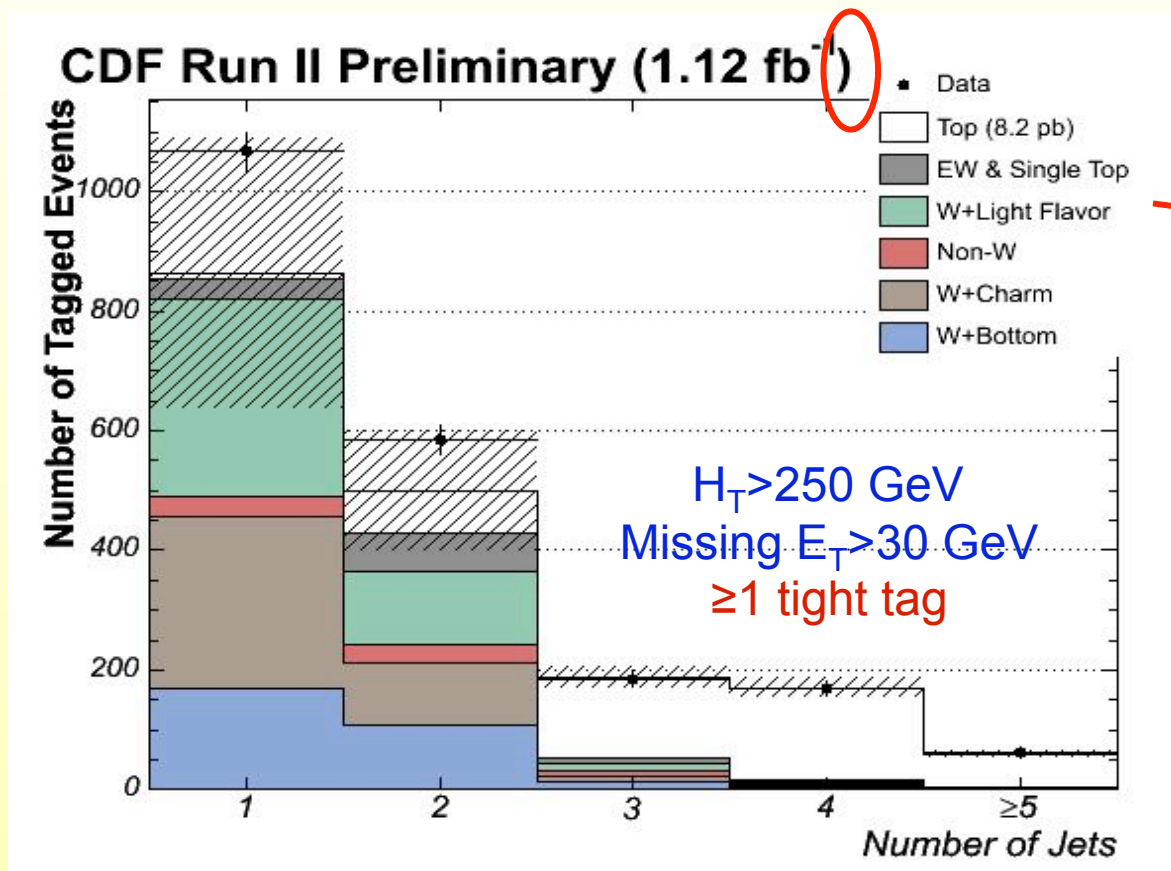


Expectations based on:

- MC predictions for EW and top backgrounds
- Fake tags (mistag matrix) applied to light jets
- Heavy Flavor background:
admixture of MC and data driven procedure



... and ... in Top analysis



Expectations based on:

- MC predictions for EW and top backgrounds
- Fake tags (mistag matrix) applied to light jets
- **CDF Method 2** admixture of MC and data driven procedure

$$\sigma_{tt} = 8.2 \pm 0.5 \pm 1.0 \text{ pb}$$

... how is this expectation obtained? →

→ HF estimates: CDF method 2

- **Method 2** = Heavy flavor background estimate in the b-tagged W + jets sample
 - Signal = $t\bar{t}$, single top, WH

- **Main backgrounds**

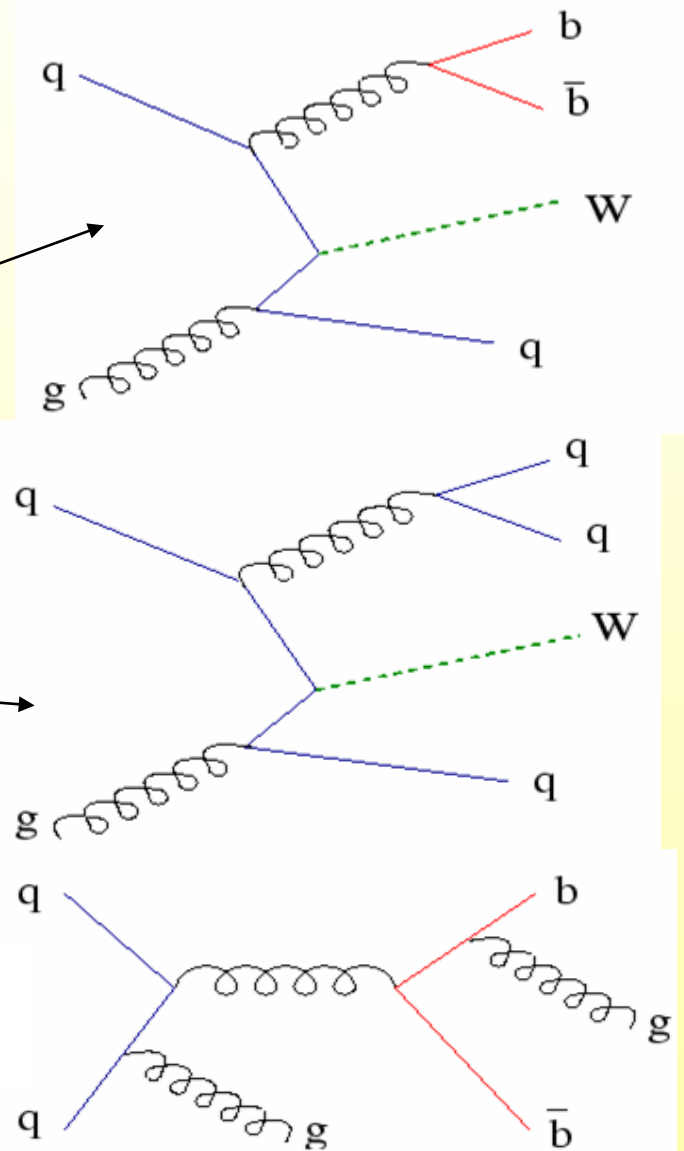
- W + HF (W+bb, W+cc, W+c)

MC (HF fractions) and Data

Data based

Data based (diboson, Z+jets, etc.)

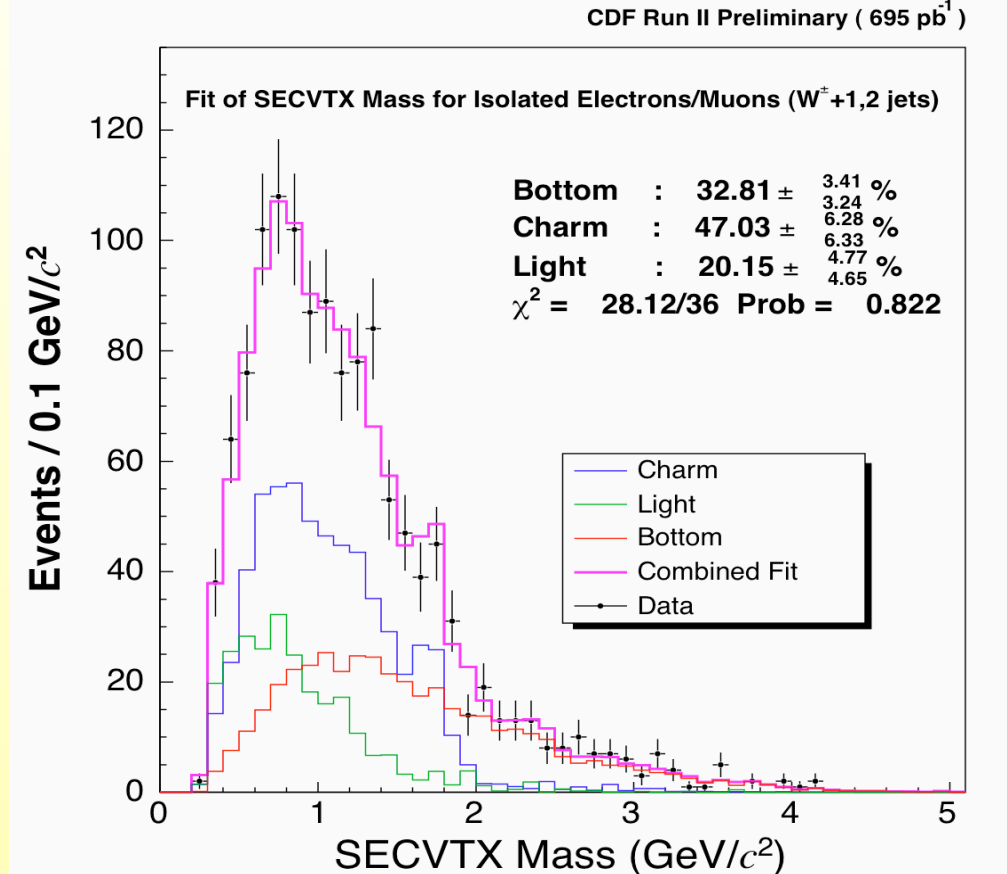
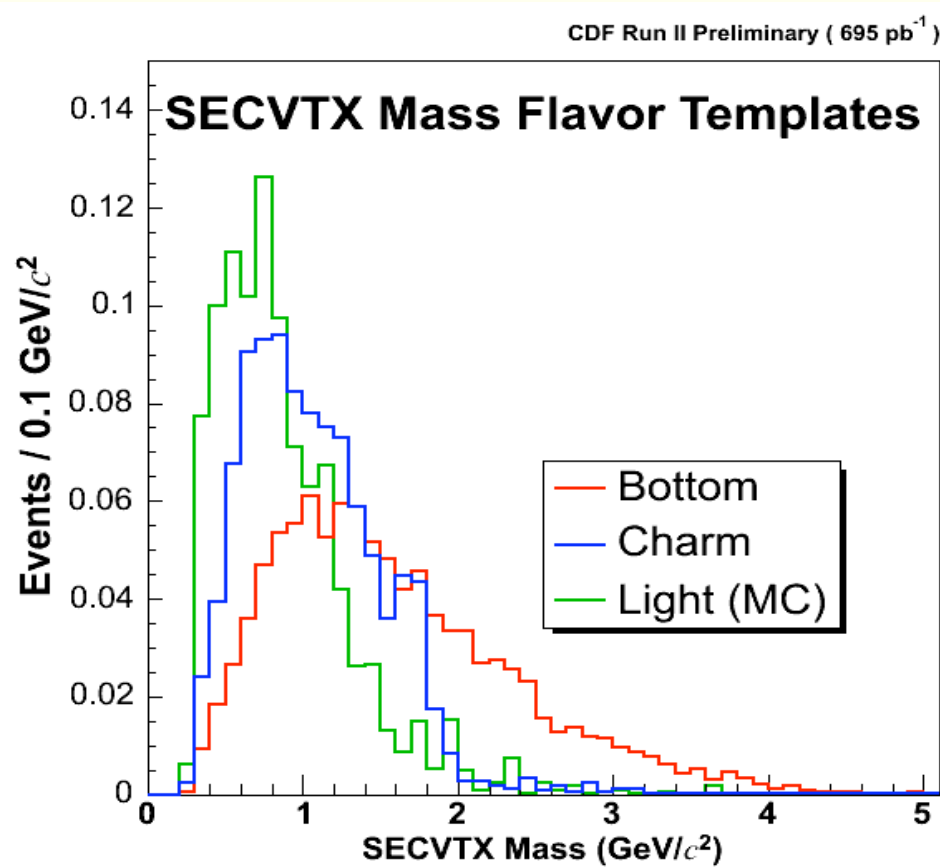
MC based



-
- *At what level do we know W/Z + HF production ?*
 - *Present measurements at Tevatron*
 - *On going studies and perspectives*

W + bb production results

In secondary-vertex-tagged sample, fit for light, c, b contributions

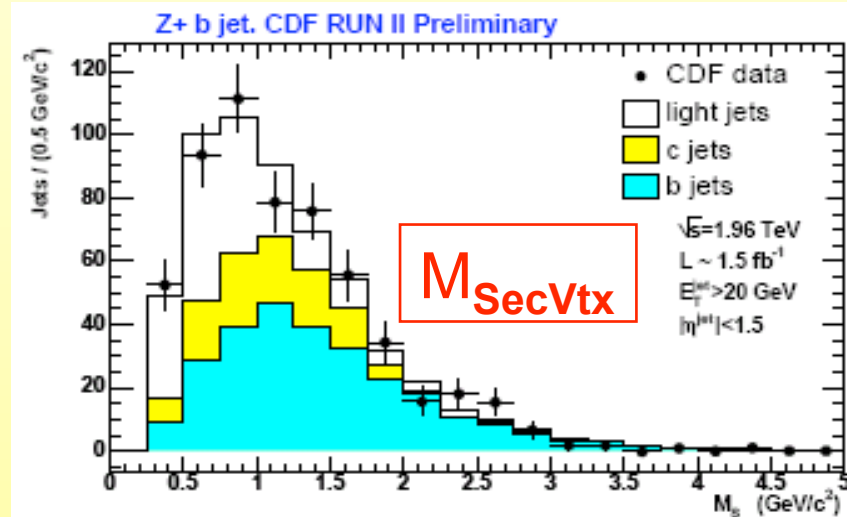
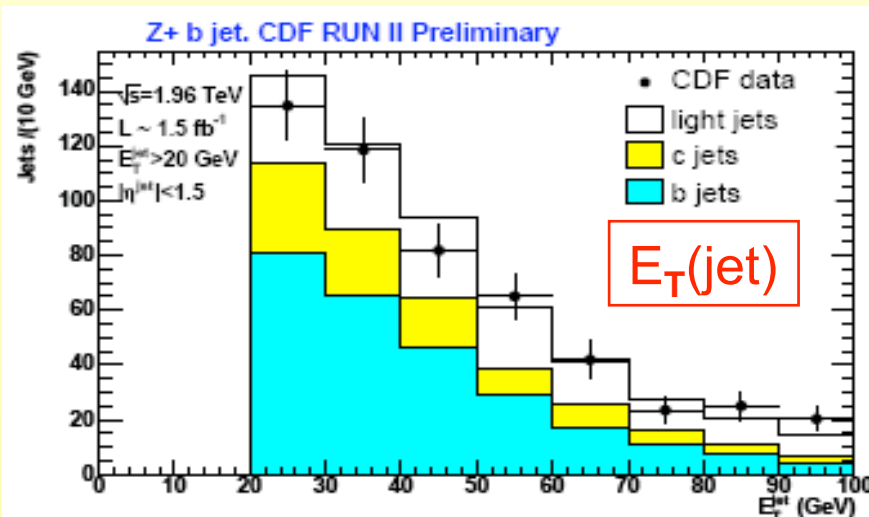
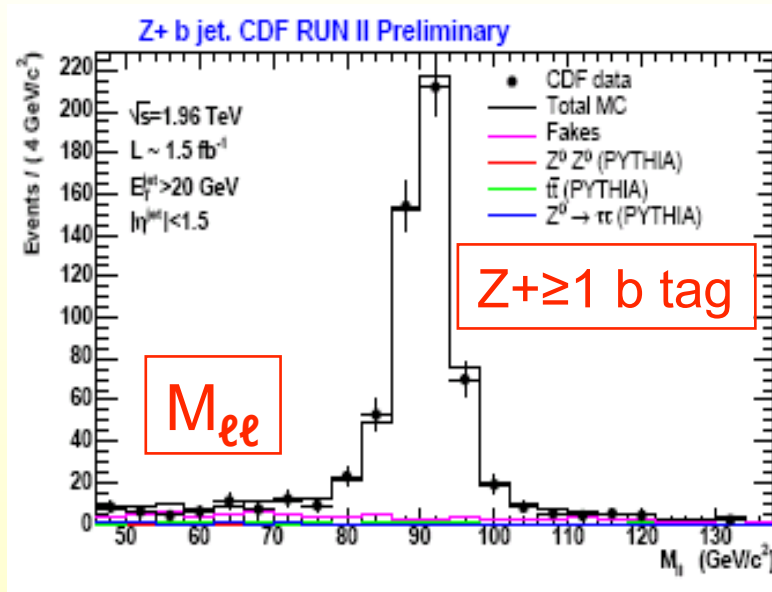
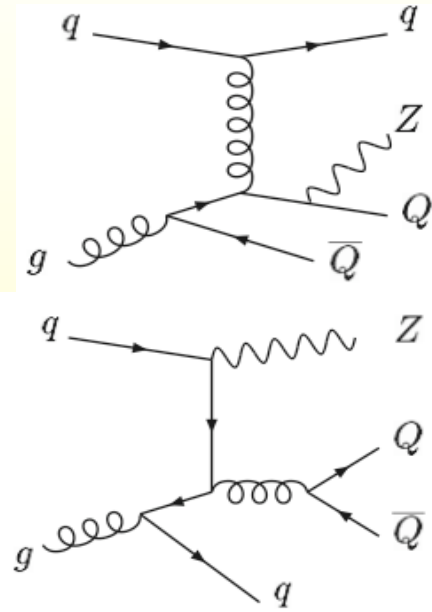


measure: 0.90 ± 0.32 pb for Wbb ($E_T(j) > 20$ GeV, $|\eta| < 2.0$)

LO calculation (ALPGEN): 0.74 ± 0.18 pb

Z+b jets analysis

Ask for the Jet to have a
Secondary Vertex Tag and
...exploit light, charm to b
separation from sec.vtx.
Mass



Z + HF x-section from Tevatron to LHC

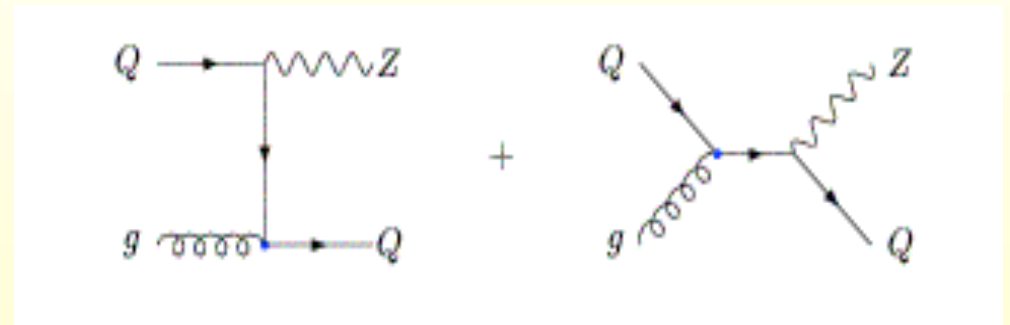
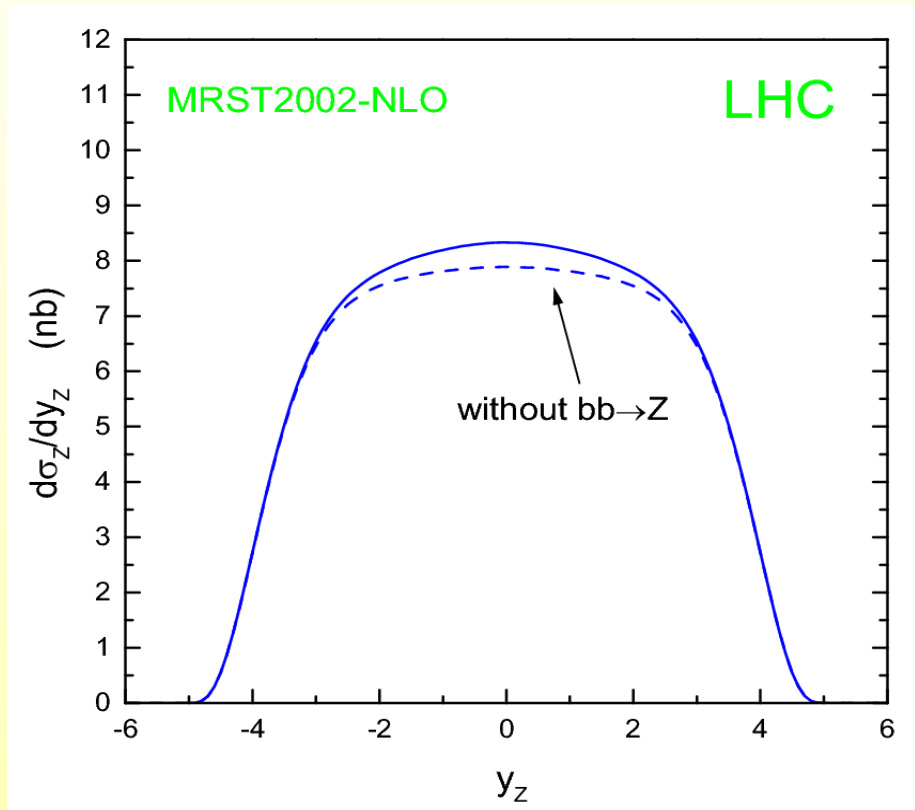
from hep-ph/0312024 - Tevatron : $p_T \geq 15 \text{ GeV}$, $|\eta| < 2$, $\Delta R_{jj} > 0.7$

Tevatron

Production (pb)	ZQ	$Z(Q\bar{Q})$	ZQj	$ZQ\bar{Q}$	Zj	Zjj
$gb \rightarrow Zb$	10.4	0.169	2.19	0.631	-	-
$q\bar{q} \rightarrow Zb\bar{b}$	3.32	1.92	-	1.59	-	-
$gc \rightarrow Zc$	16.5	0.130	3.22	0.49	-	-
$q\bar{q} \rightarrow Zc\bar{c}$	5.66	6.45	-	1.70	-	-
$q\bar{q} \rightarrow Zg, gq \rightarrow Zq$	-	-	-	-	870	137

LHC
~1000
~50
~1400
~90
~16000

LHC example: b quark PDF from $Z+b$



$bb \rightarrow Z$ @ LHC is $\sim 5\%$ of entire Z production

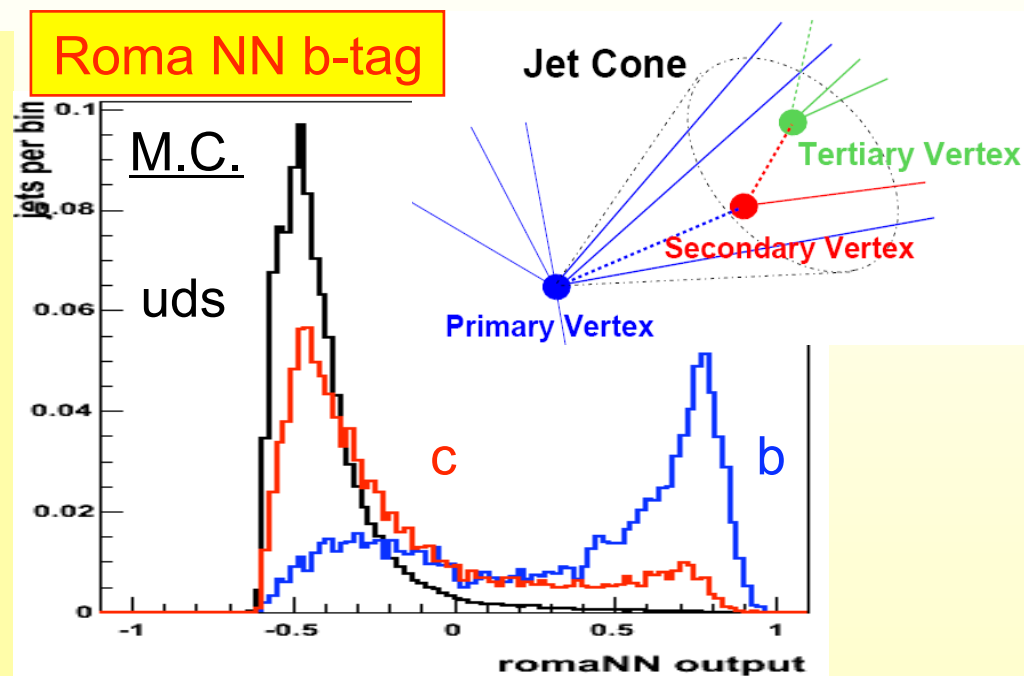
Spread of existing pdf's gives a 10% spread in the prediction of the SM Higgs cross section

Even bigger impact on MSSM Higgs production (sensitive to bb coupling)

-
- *New generation of Jet Flavor tagger*
 - *Highlights on first attempt to measure directly:
Z+b , Z+c , Z+uds*
 - *go beyond method 2 like analysis*

A new CDF H.F. Neural Network Tagger

- The underlying idea is improving the ability to boost performance exploiting *as much information as possible*:
 - number of vertices (not just **A SINGLE** secondary vertex a' la SecVtx) and their masses;
 - displaced but un-vertexed tracks (tracks not belonging neither to secondary vertex nor to the primary);
 - Soft leptons, JetProb, global jet variables
- Best to combine all info together in a single discriminant: use (a series of) *Neural Networks* (NN)
- NN training performed on generic sample to be used in different analysis (top, single top, low mass Higgs, ...)

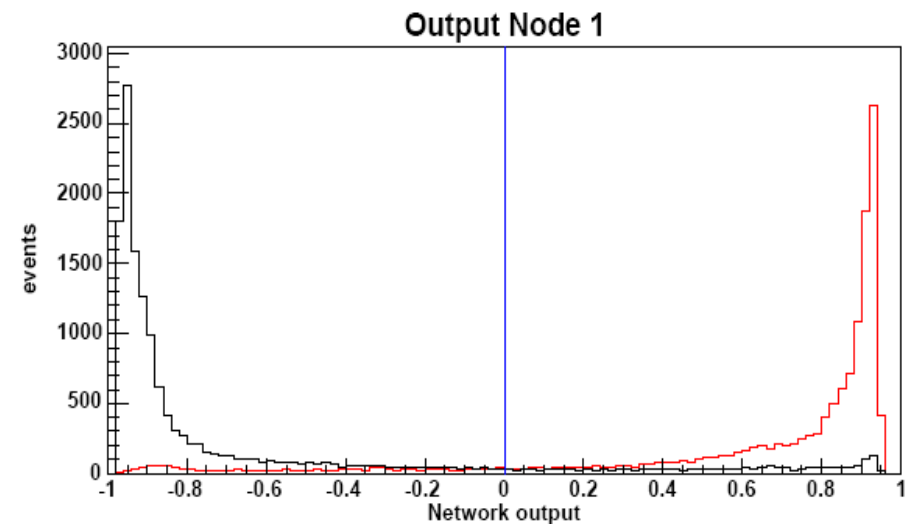
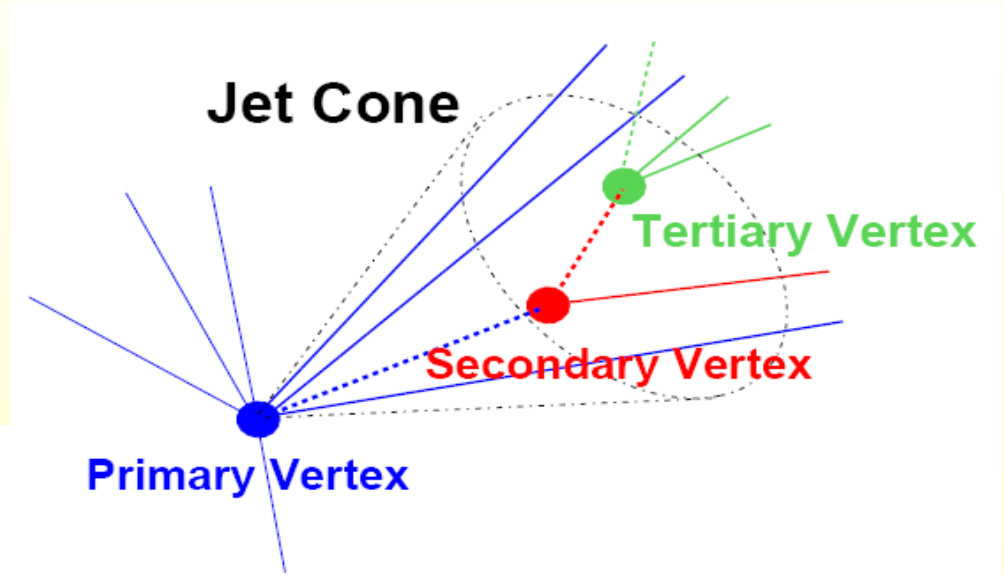


- **A Continuous tagging** variable allow weighting events to get the most of it in terms of search sensitivity
- **GOOD b-c-uds statistical separation** allows the fraction of c jet to be studied as well, HIGH Jet taggability: $\sim 75\%$
- **Increased per-jet efficiency** at same background rate than SecVtx:

1st step: new vertexing algorithm

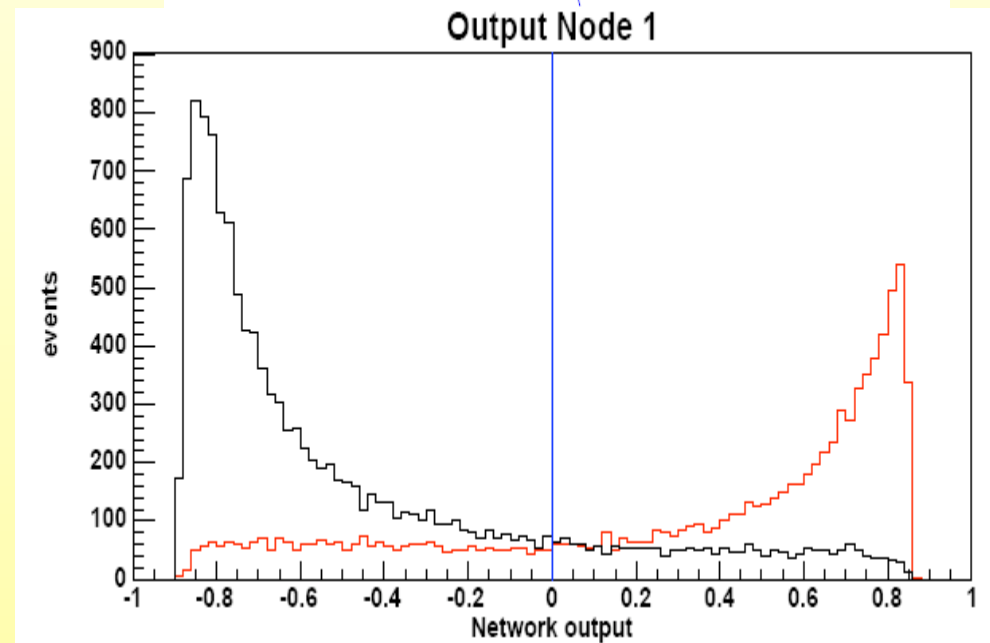
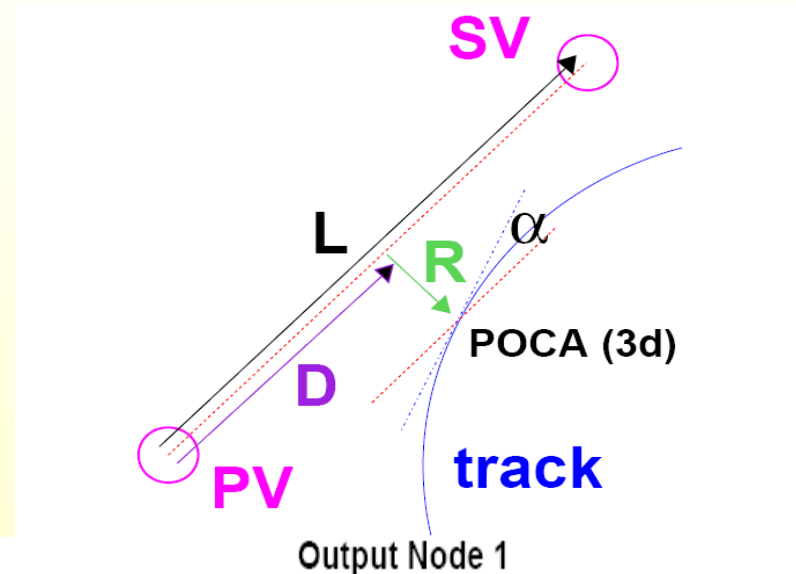
Uses all tracks in the jet to try to find **multiple vertices**: primary+all secondaries. No cut on decay length as usual b-tags.

- b-hadrons often produces more than one distinct vertex
- Jet containing two b-hadrons have even more
- Tracks are unambiguously assigned to “closest” vertex based on their χ^2
- Iterative procedure ends when no more track can be associated to any vertex
- List of vertices found and associated tracks passed to next stage of NN (un-vertexed tracks and HF discrimination)



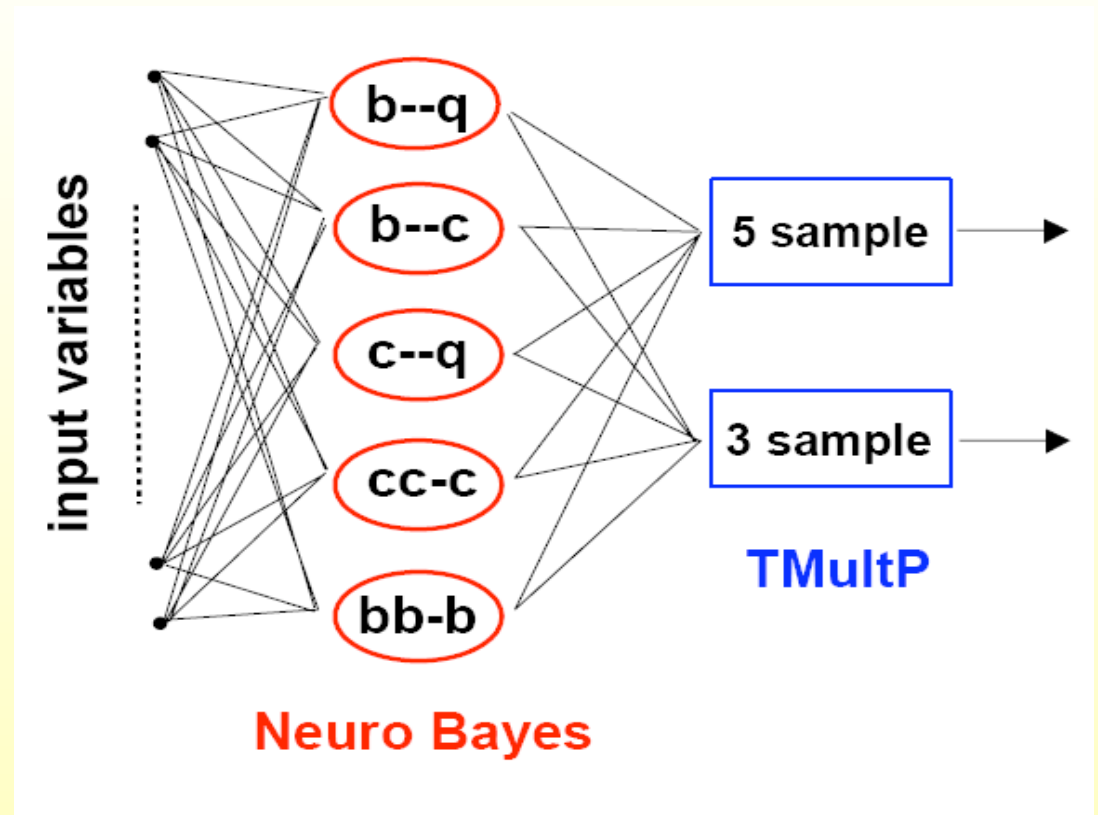
2nd step: track NN

- Selects **un-vertexed** tracks from HF decays
- Select 5 most significant variables and train network to distinguish:
- good **track-vertex combinations** (true HF decay track and true HF vertex decay)
- from other combinations (fake track-fake vertex, fake-signal etc.)



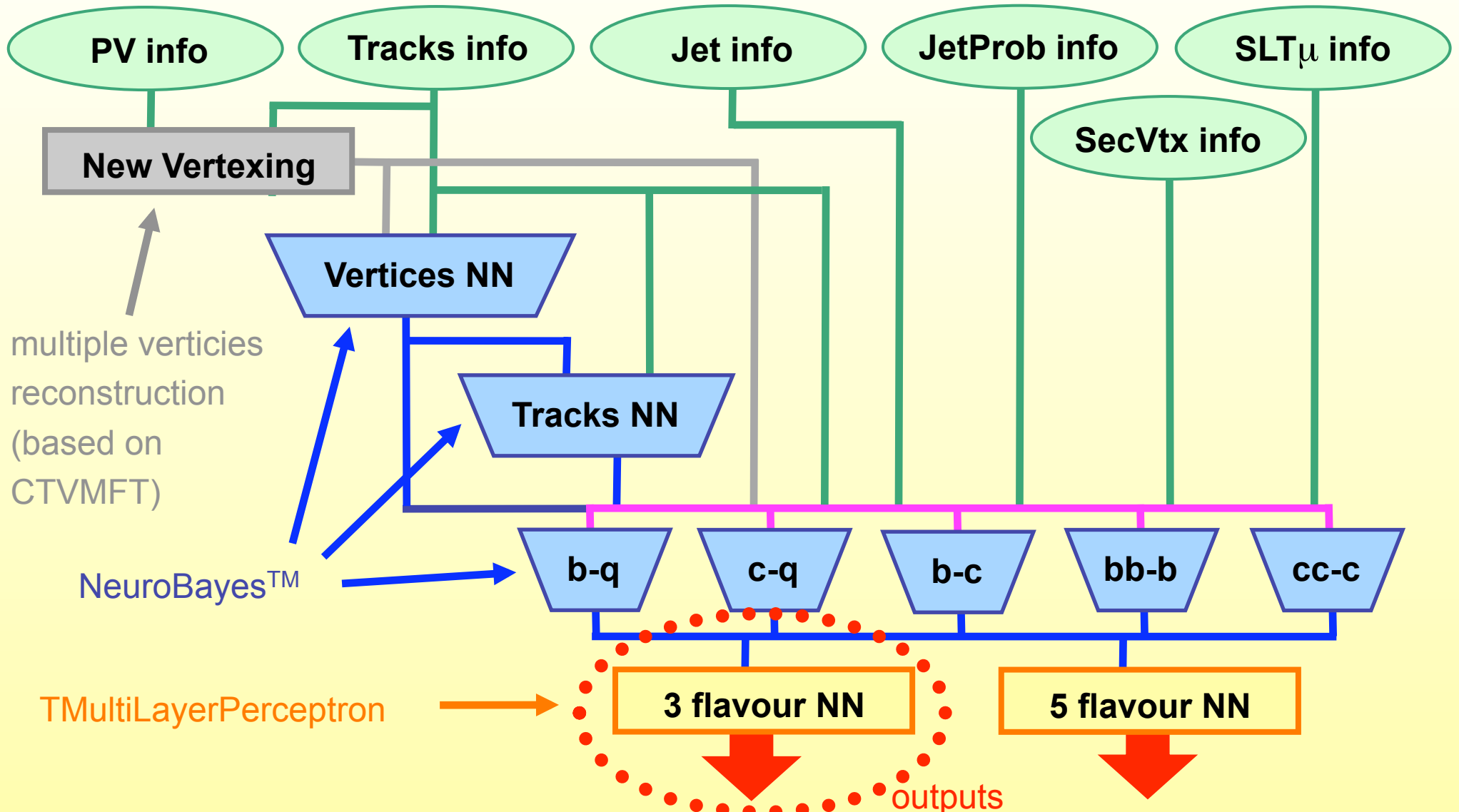
3rd step: Jet NN

- Combine tracks and vertices NN outputs, muons, SecVtx, JetProb, etc.
- 5 different NN's specialized to separate two by two light, c, b, ccbar, bbbar jets



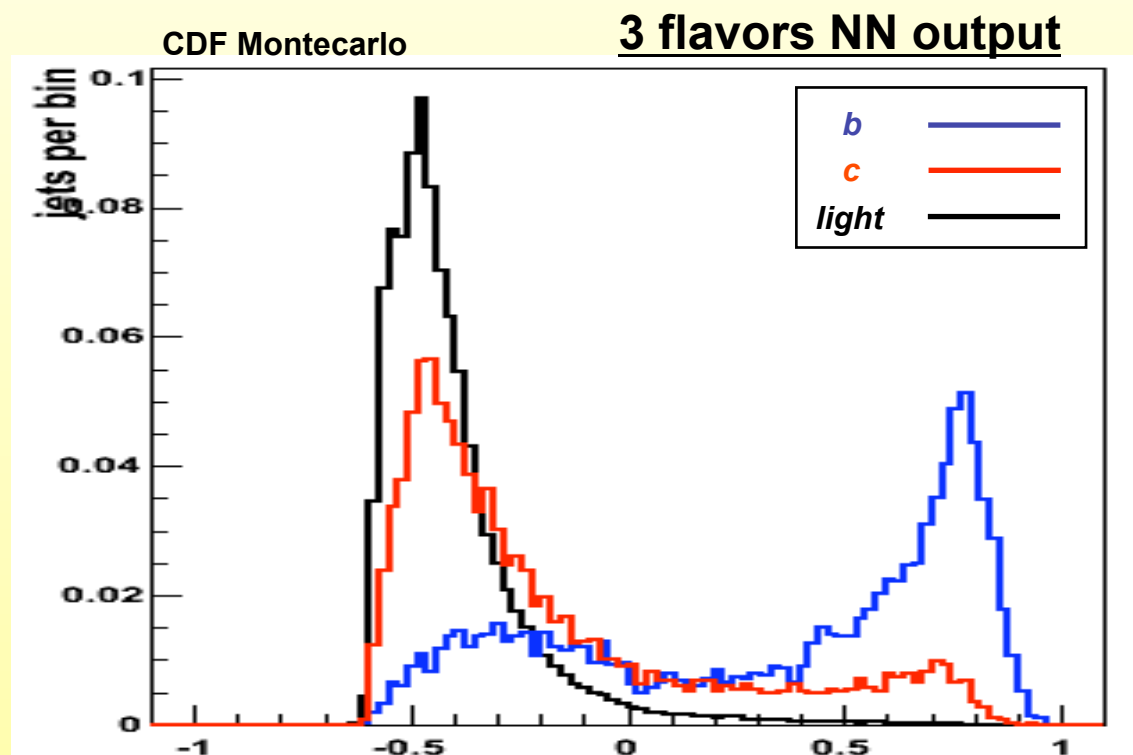
- Combine the 5 outputs to give a 5 flavour separation (light, c, ccbar, b, bbbar) or 3 flavour (light, c, b) in a single tagging variable
- Present studies just on the simpler **3 flavors NN final output**

NN tagger structure



Roma NN final output

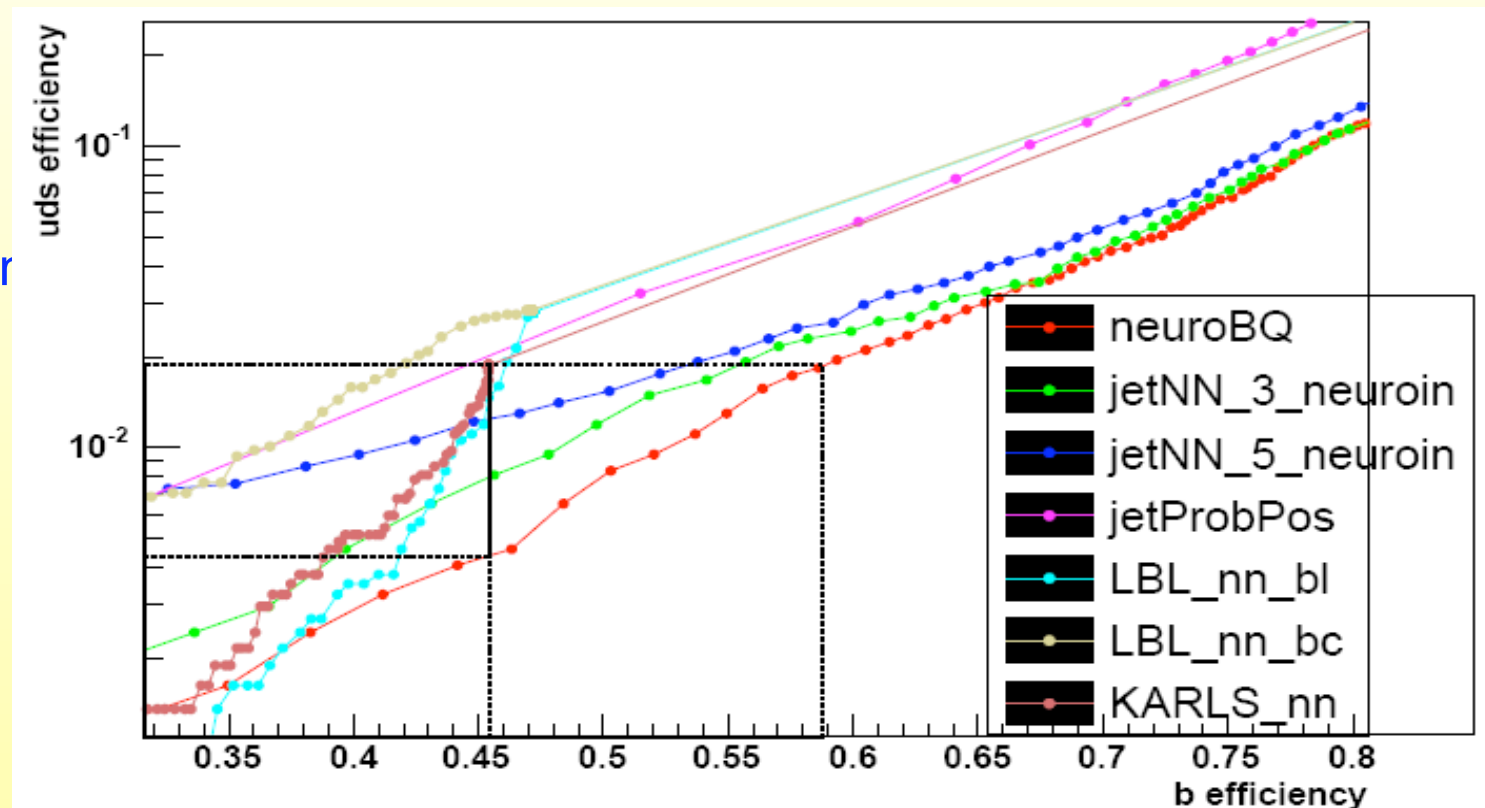
- A **Continuous** tagging variable allow weighting events to get the most of it in terms of search sensitivity
- **GOOD *b-c-uds statistical separation*** allows the fraction of c jet to be more reliably studied.
- **HIGH *Jet taggability***: $\sim 75\%$



Advantages

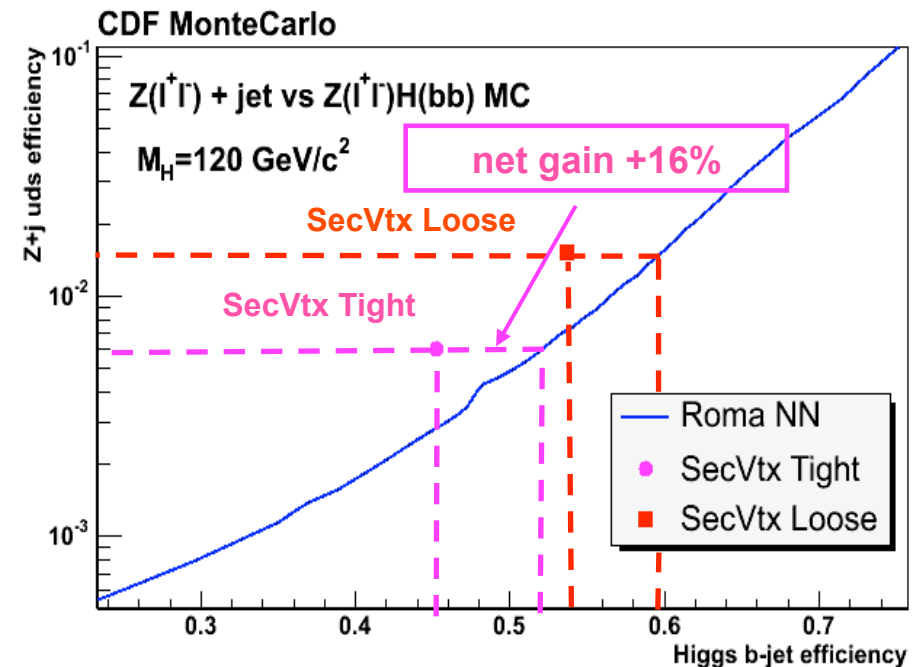
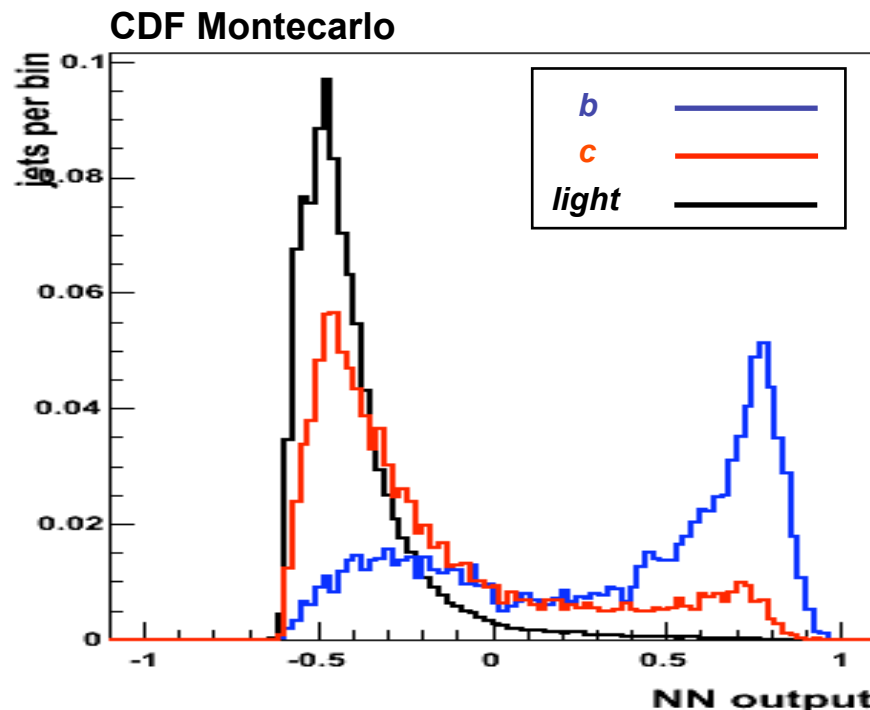
- Work in pre-tag sample (Roma NN Jet taggability $\sim 75\%$)
- Classify jets according to purity
- More efficiency at same background rate than SecVtx

- Similar or better performances than Karlsruhe and Berkeley Neural Networks (proved on MC)



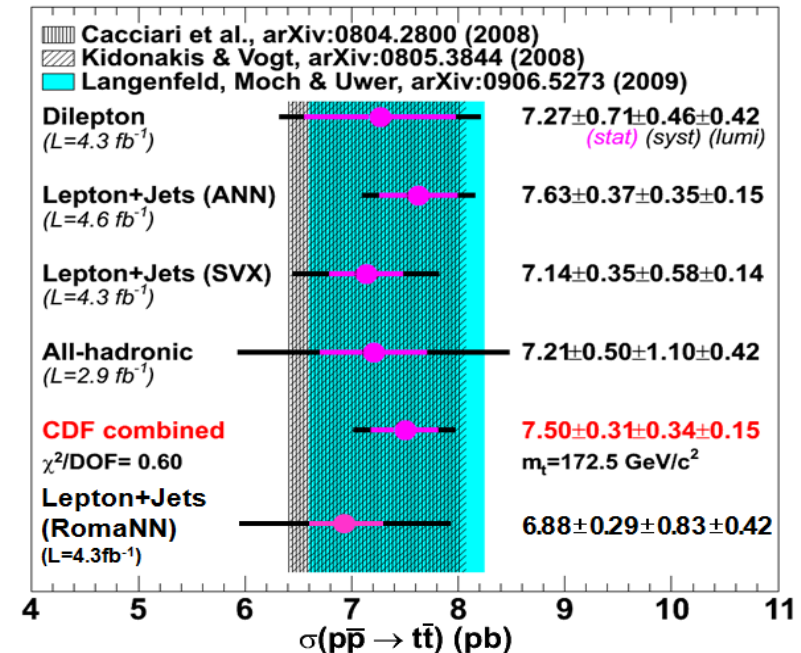
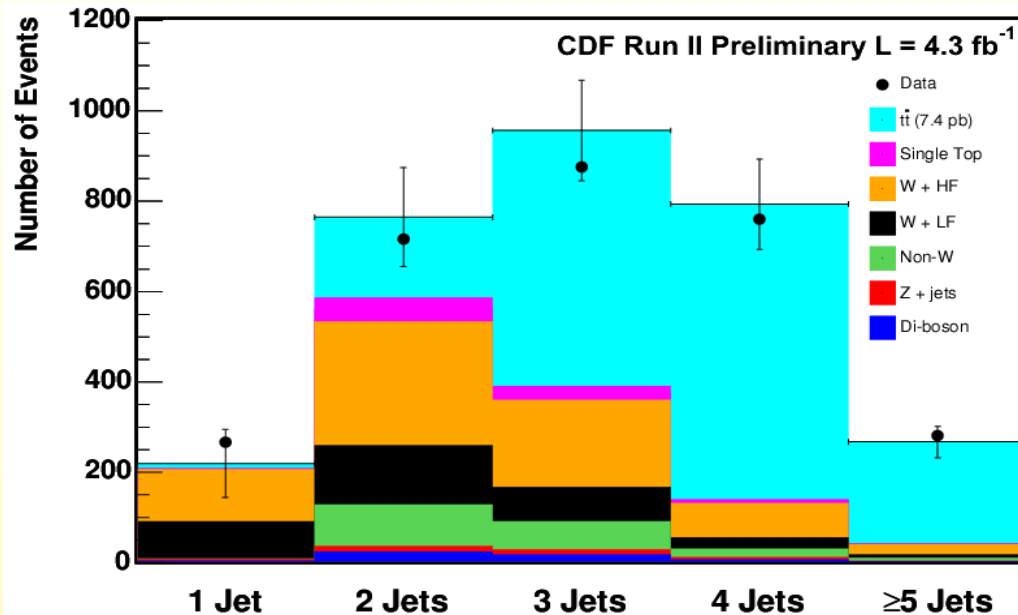
Advantages

- Increased **per-jet efficiency** at same background rate than SecVtx:
 - +30% (relative) on a $p\text{-}\bar{p} \rightarrow q\text{-}\bar{q}$ MC di-jet (Jet 20 - btopqb)
 - +16% (relative) for Z+j compared to ZH(120) MC



Top x -section in Lepton+Jets with cut based Roma NN b -tag

New !

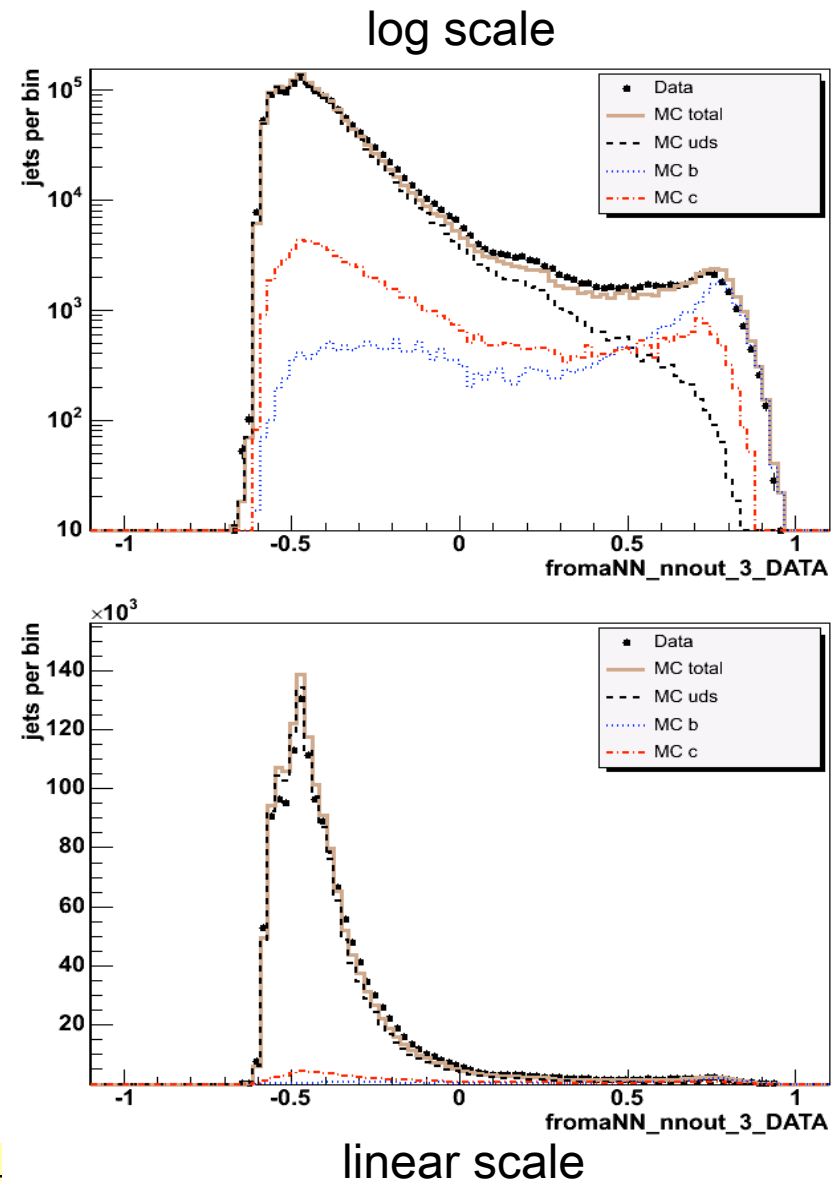


- The cross section of pair produced top quarks has been measured to be: $6.88 \pm 0.29_{\text{stat}} \pm 0.83_{\text{syst}} \pm 0.42_{\text{lumi}}$
- using 4.3 fb^{-1} of collected data from the high Pt lepton triggers. The measurement is performed with TightRomaNN tagged lepton plus jets events with ≥ 3 tight jets, $H_t \geq 230 \text{ GeV}$, and $\text{MET} \geq 20 \text{ GeV}$. This measurement has taken the top mass to be $172.5 \text{ GeV}/c^2$.

Higgs search in WH \rightarrow In bb with RomaNN in the pipeline ...

Tuning NN output to data (aka calibration)

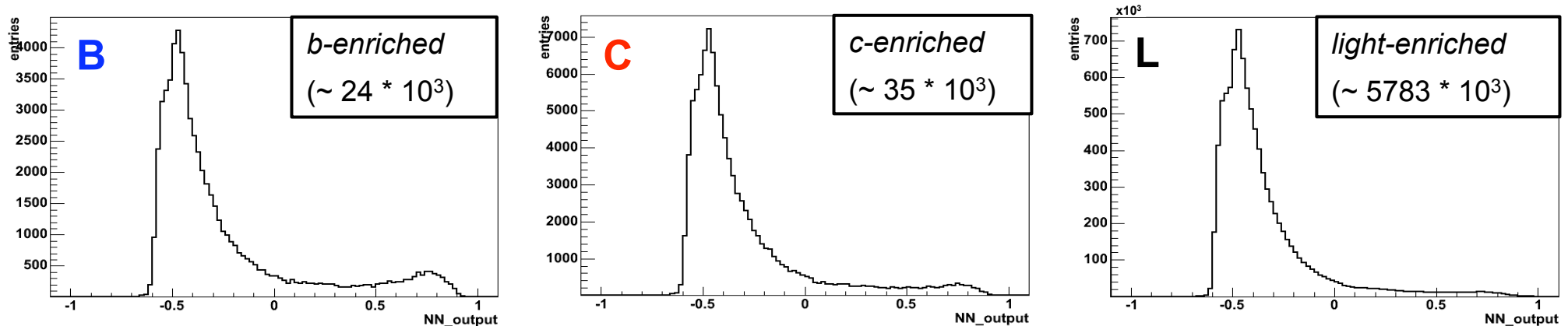
- iterative procedure to extract **correction functions** (cf) = weights for b, c and light MC tagger templates
- **Data driven** extraction of the cf's exploiting the high statistic JET20, JET50 data samples and the corresponding di-jet MC samples
- Defined 3 independent and unbiased jet samples with different HF content but otherwise identical jet properties
 - each flavor has the same shape in all the 3 samples
- Correct for Data/MC disagreement & measure HF content at the same time



MC(Pythia) predicted b/c/light fractions

Calibration data-set

- Define 3 independent flavor enriched samples selecting di-jet events and asking for an *away jet*:
 - SecVtx tag and b/c LBL NN > 0.6 *b-enriched* (B)
 - SecVtx tag and b/c LBL NN < 0.6 *c-enriched* (C)
 - SecVtx untagged *light-enriched* (L)

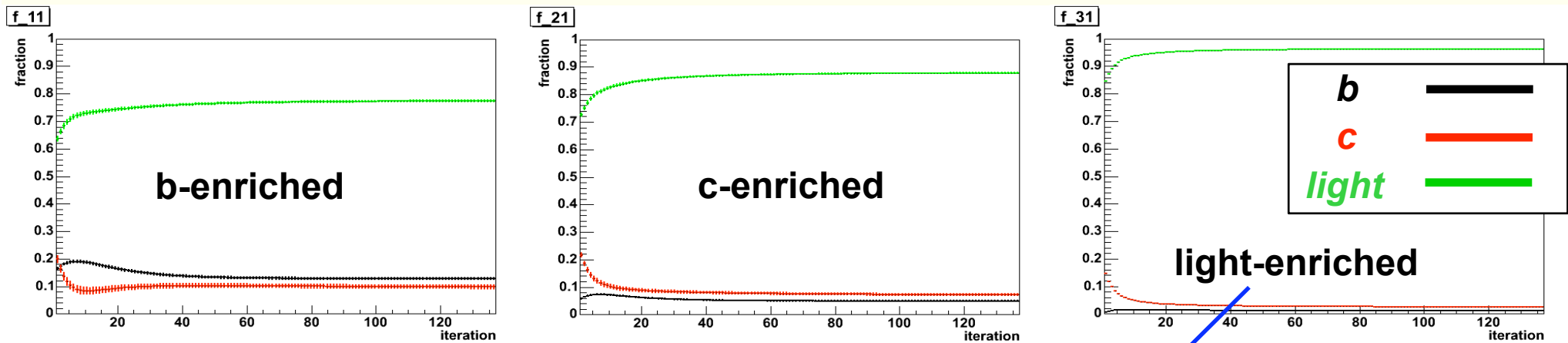


MC prevision

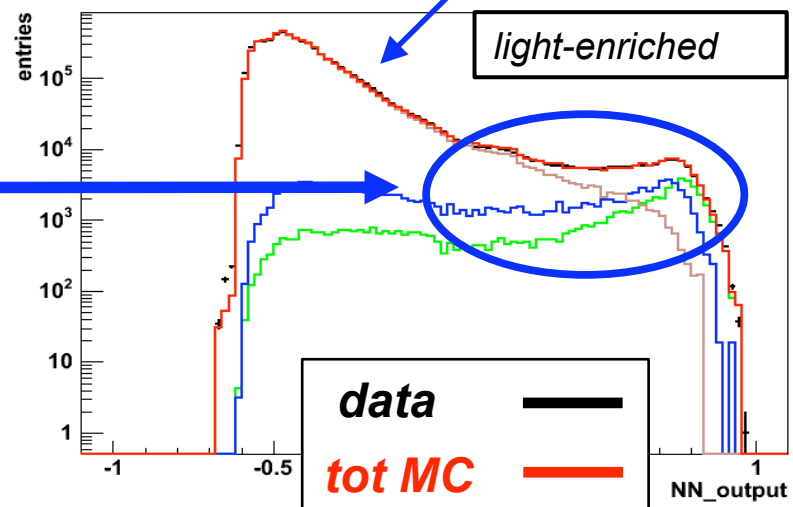
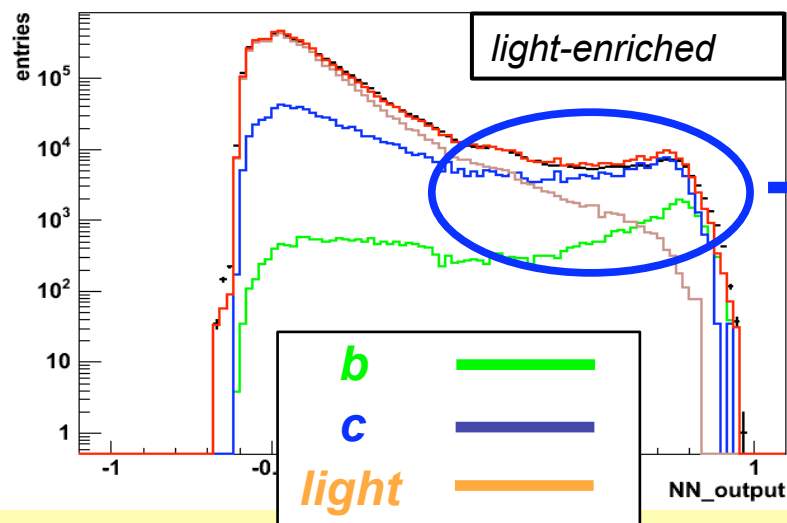
Sample	b-jets	c-jets	light-jets
B	0.234 ± 0.009	0.058 ± 0.004	0.708
C	0.096 ± 0.005	0.092 ± 0.004	0.812
L	0.022	0.052	0.926

Templates calibration with Jet DATA

Fraction evolution as a function of the iteration.



before correction

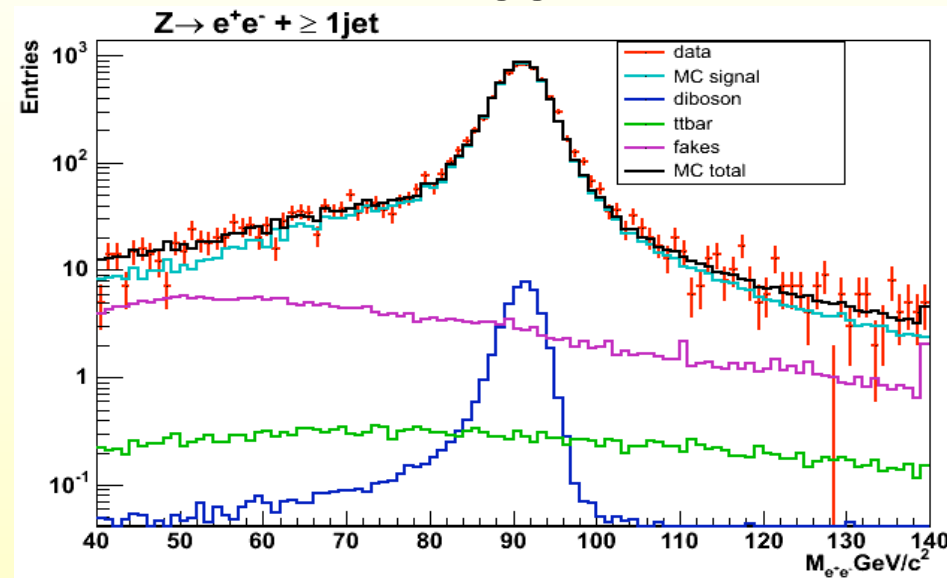


after correction

Roma NN full application: Z+HF

- Ideal play-ground to test the new Roma NN tagger

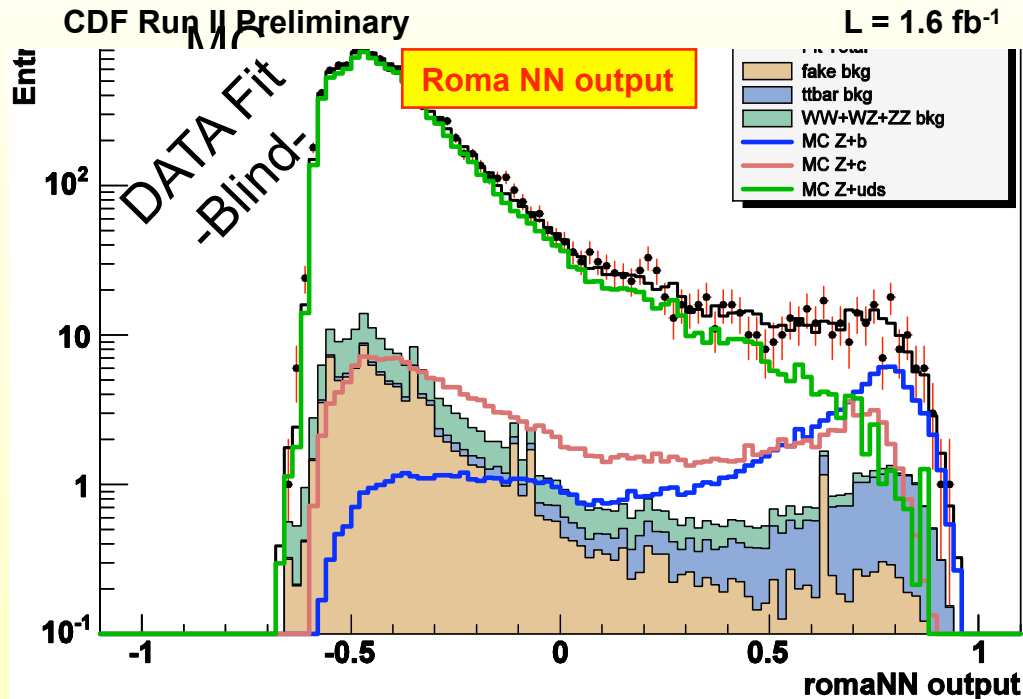
- Easy signature
- Low jet multiplicity
- Analysis based on 1.1 fb^{-1}



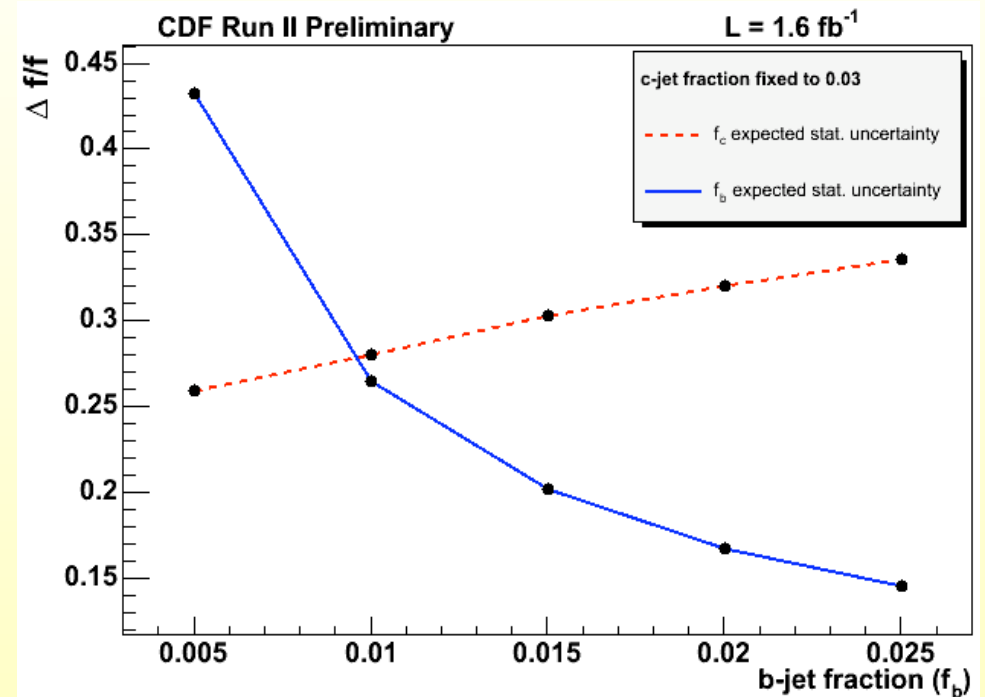
- Interesting measurement by itself: QCD generators, PDF's...
- Major background to ZH channel Higgs search
- Understand how to extrapolate corrected templates from generic di-jet samples to Z+jets signal sample:
 - STRATEGY: re-weight di-jet data to match E_T spectra in the target sample (eventually re-weight for other event related observables)
 - Test case for other extrapolations: to top, Higgs,...

Z+HF fractions with Roma NN b-tagging

Roma NN b-tag output fit in Z+jets events, $L = 1.6 \text{ fb}^{-1}$



Toy MC estimate on b , c fractions uncertainty

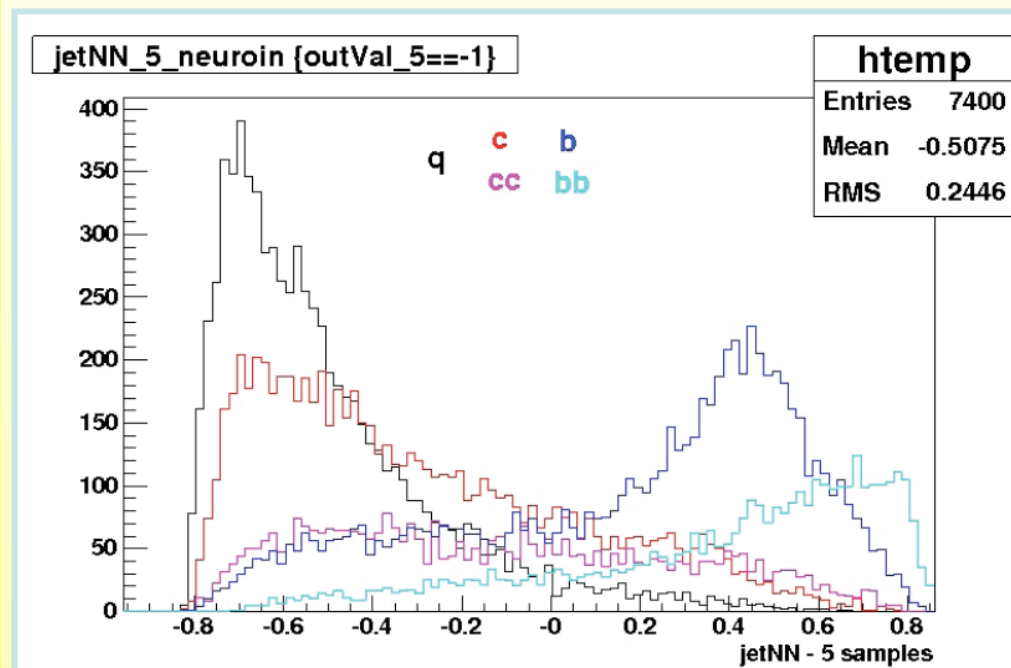


- Could allow for the first time a simultaneous measurement of $Z+c$ and $Z+b$ cross sections.
- Application of the Roma NN tagger to other samples: one could fit top/higgs cross section together with the W/Z+HF background as a function of jet multiplicity.

New generation Jet Flavor Tagger, Outlook

- The whole concept of Method II could be revisited also with the use of a powerful, high efficiency and well simulated b-tagging tool → **change paradigm**:
- No need of working out the tagged sample from the pretag sample and questionable HF fraction from MC
- In principle one could fit the top/higgs cross section and the $V+b/c$ component in each jet multiplicity bin taking in to account fake, and other EWK bkg! ... *to be followed up ... and discussed...*

HF sample composition for Higgs, Top and searches analysis is becoming more and more crucial

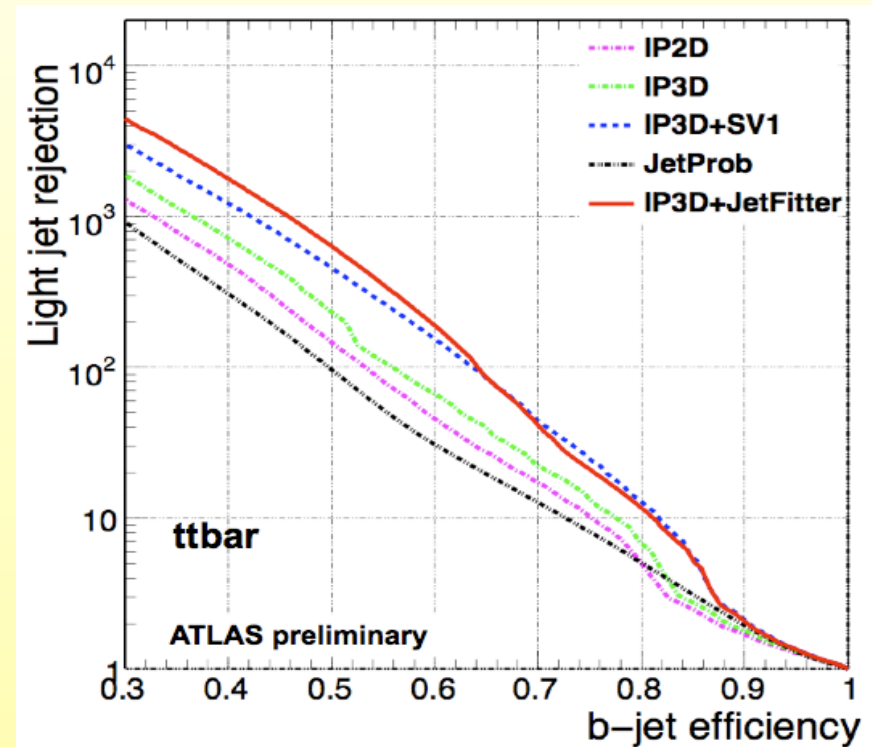
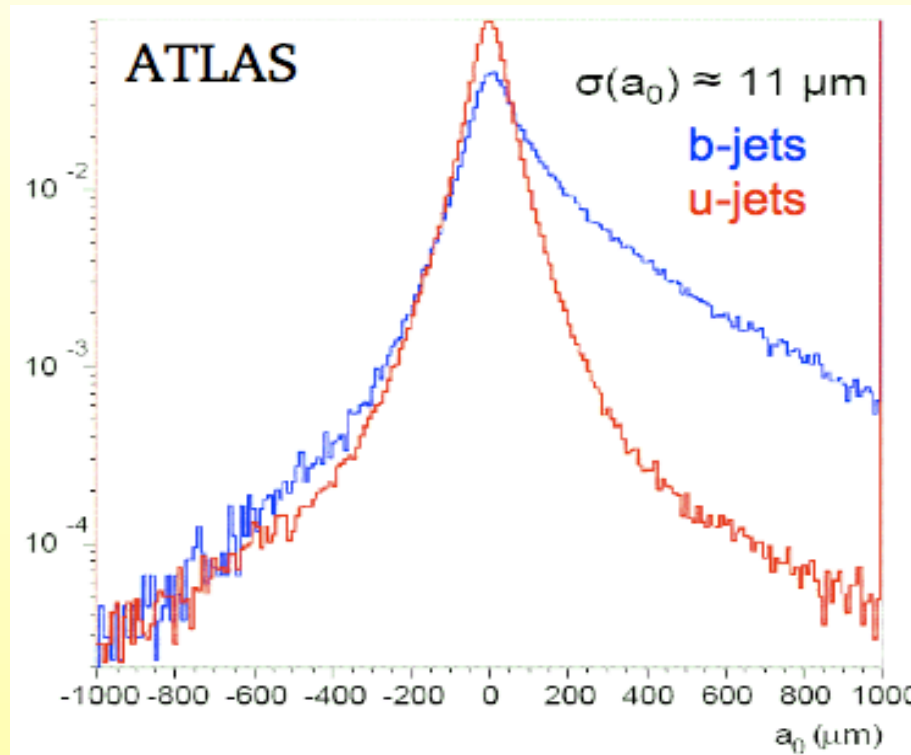


- **Further possibility:** building block are there to tackle also the b/\bar{b} (c/\bar{c}) jet issues

-
- *a look to LHC now*

B-tagging in ATLAS

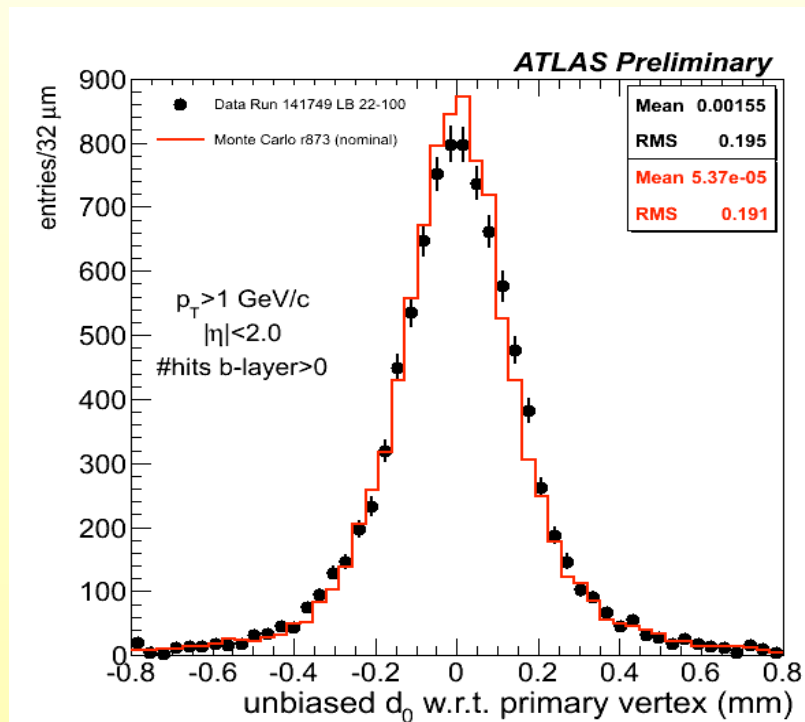
- Many studies has been made for soft-leptons, IP tags and multivariate
- Also b-Jet trigger at HLT level based on likelihood
- For now concentrate on IP tag validation of performances, **on MonteCarlo:**



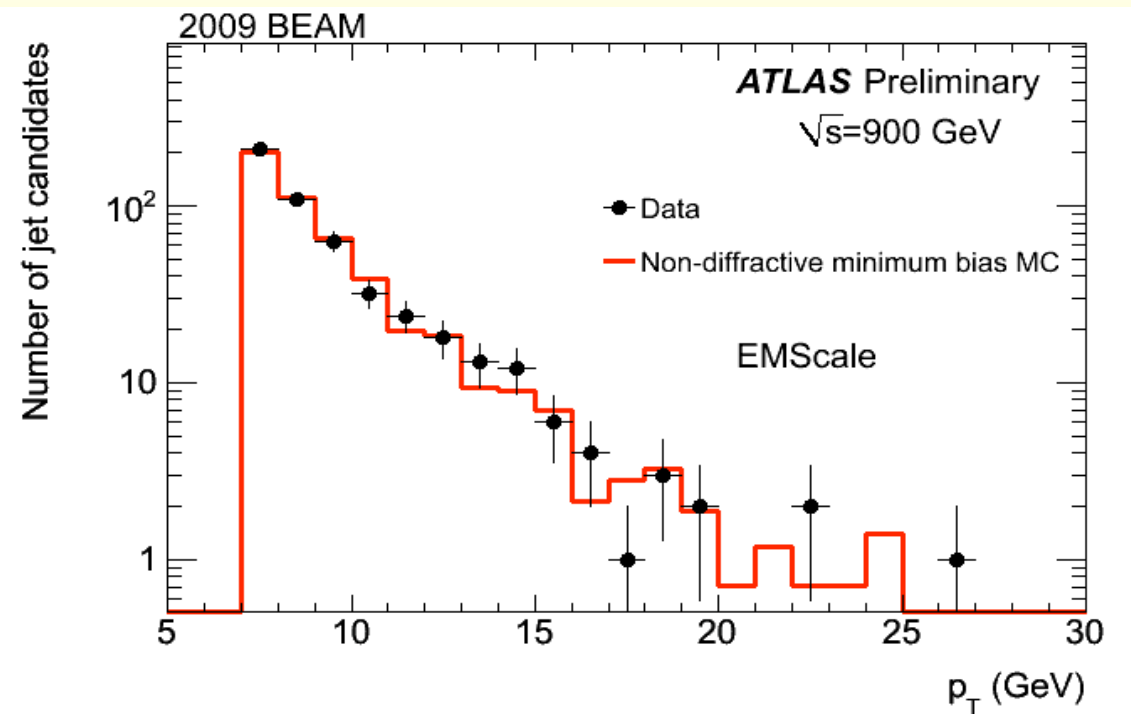
B-tagging in ATLAS

- For the first data concentrate on IP tag validation ... and on DATA:

Tracks Impact parameter



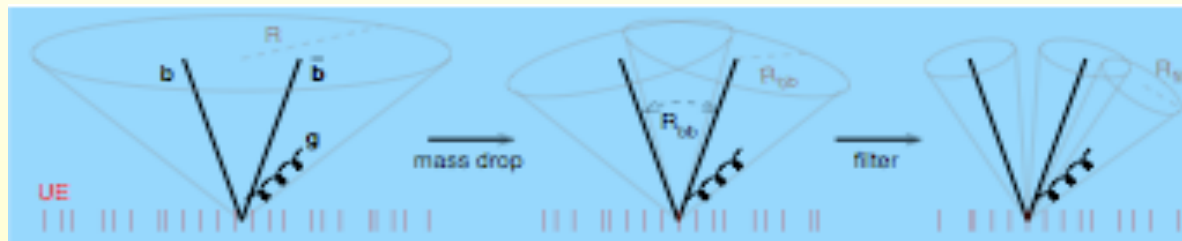
Anti-KT cone 0.4 Jets



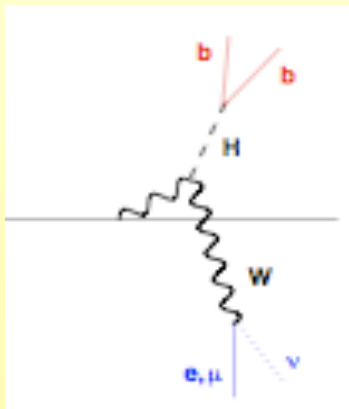
Future: boosted objects decaying in bb ?

At LHC one can have highly boosted sizeable fraction of heavy particles decaying in bb and merging in single reconstructed JET:

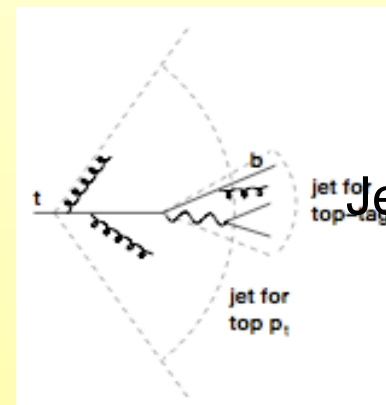
- try to explore Jet substructure, guided by b-tagging interest regions



Possible applications to:



WH with $H \rightarrow bb$ or



Jet "top – tagging"

Summary:

- Reviewed main b-tagging strategies and their use
- Highlight the relevance of b-tag control for background processes studies
- W/Z + HF processes not yet fully studied, now starting to characterize at the Tevatron.
- New generation NN Jet Flavor Tagger being developed at CDF and being tested in Z+b, Z+c measurement and used in Top and Higgs analysis
- *... lot of knowledge progresses on HF production and b-tag are an important Tevatron legacy to the LHC ... a rich research ground for next years...*
- B-tagging is a fun and challenging topic
 - Design/calibration of taggers has a lot of physics in it
 - Not a solved problem – still room for good ideas

Thank you

Backup

Method 2 on One Slide

- Start from the W+jets data before tagging (pretag)

- Estimate MC-based backgrounds and Non-W

$$N_{MC} = \left(\sum_i \sigma_i \varepsilon_i \right) \int \mathcal{L} dt \quad N_{NonW} = f_{nonW} N_{Data}^{Pretag}$$

$$N_W = N_{Data}^{Pretag} - N_{nonW} - N_{MC} - N_{signal}$$

- Subtract to get pretag W yield

$$N_{W_{HF}}^{btag} = N_W F_{HF} \varepsilon_{btag}$$

- Multiply by heavy flavor fraction and tagging efficiency to get W+bb, W+cc, W+c

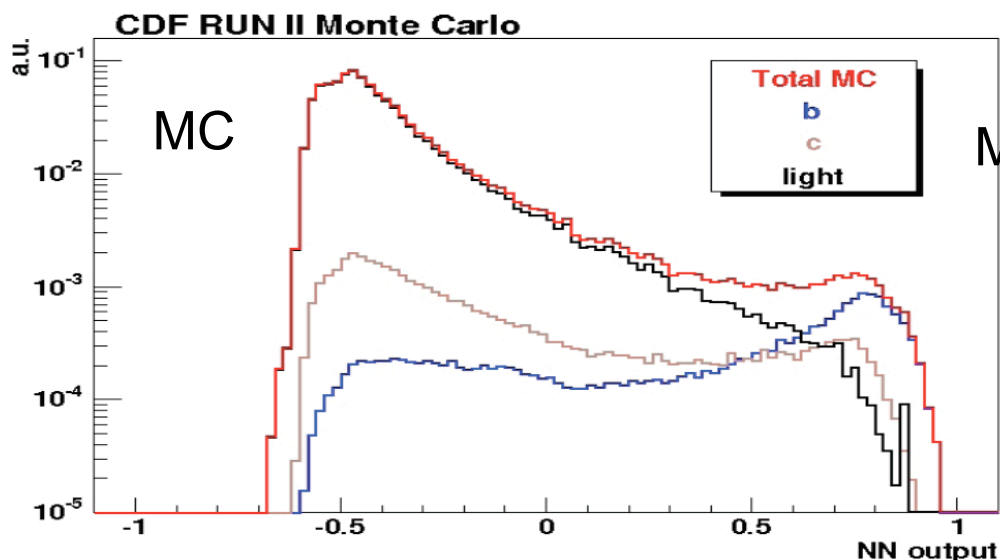
$$N_{W_{LF}}^{btag} = N_{Data}^{NegativeTag} \left(\frac{N_W - N_{W_{HF}}}{N_{Data}^{Pretag}} \right)$$

- Subtract W+HF from pretag and get W+light flavor from mistag matrix

$$N_{MC}^{btag} = \left(\sum_i \sigma_i \varepsilon_i \varepsilon_{btag} \right) \int \mathcal{L} dt \quad N_{NonW}^{btag} = f_{nonW} N_{Data}^{btag}$$

- Estimate tagged backgrounds from MC-based and non-W backgrounds

Z+HF fractions expected uncertainties

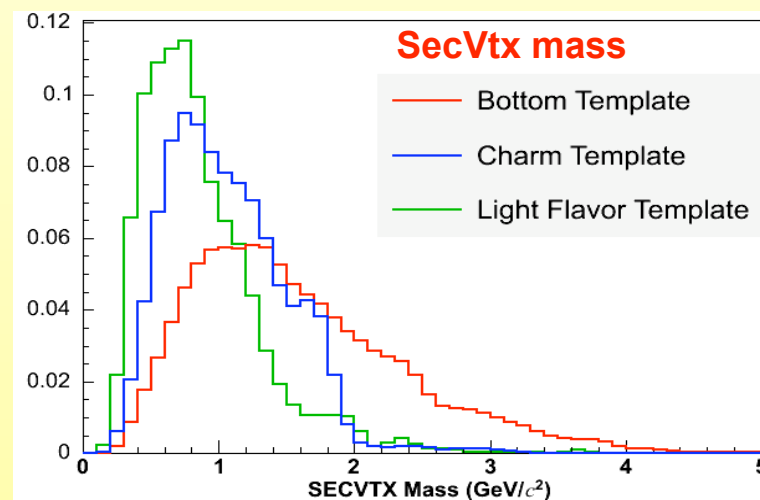


Fitted fractions with statistical error (MC)

b fraction	0.0200 ± 0.0036
c fraction	0.040 ± 0.014
light fraction	0.940 ± 0.011

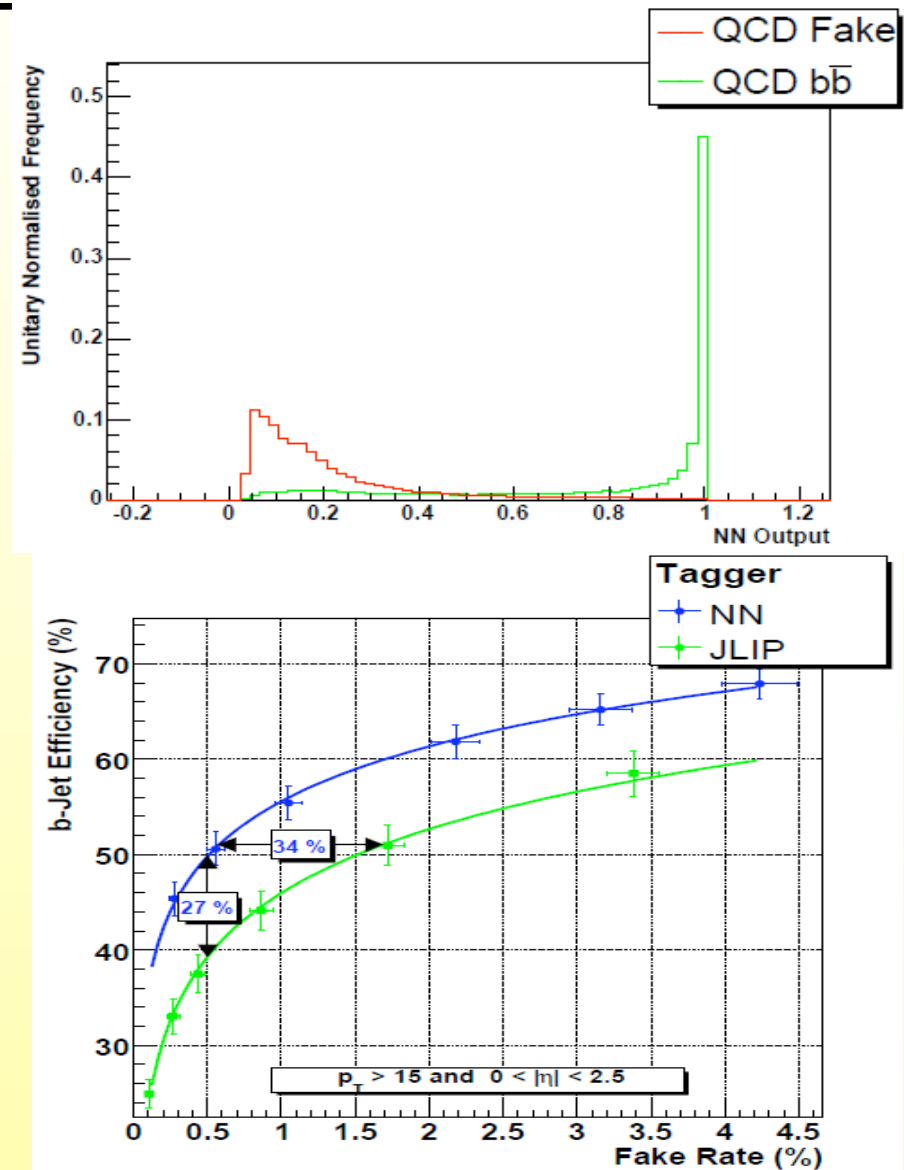
- Used MC templates to evaluate expected uncertainty
- Generate 1000 experiments with 10000 events each with input fractions of b, c, light 2,4,94 [%]:

- Allow a measurement of **Z+c** cross section with ~30% relative precision
- Statistical separation for b-jets better than **SecVtx mass** by roughly 15% for equal luminosity



Multivariate Tagger (DØ)

- 7-variable neural network
 - Vertex mass, L_{xy}/σ , χ^2
 - Number of vertices/jet
 - Jet total and displaced track multiplicities
 - Jet probability
- Vertex tagger uses a “super-loose” tune to get info on more jets
- Train to separate b-vs-light
- Can improve efficiency or mistag by 25%, up to a factor of two in the efficiency/purity extremes
- Efficiency and fake rates have been measured in data
- Ready for analysis use

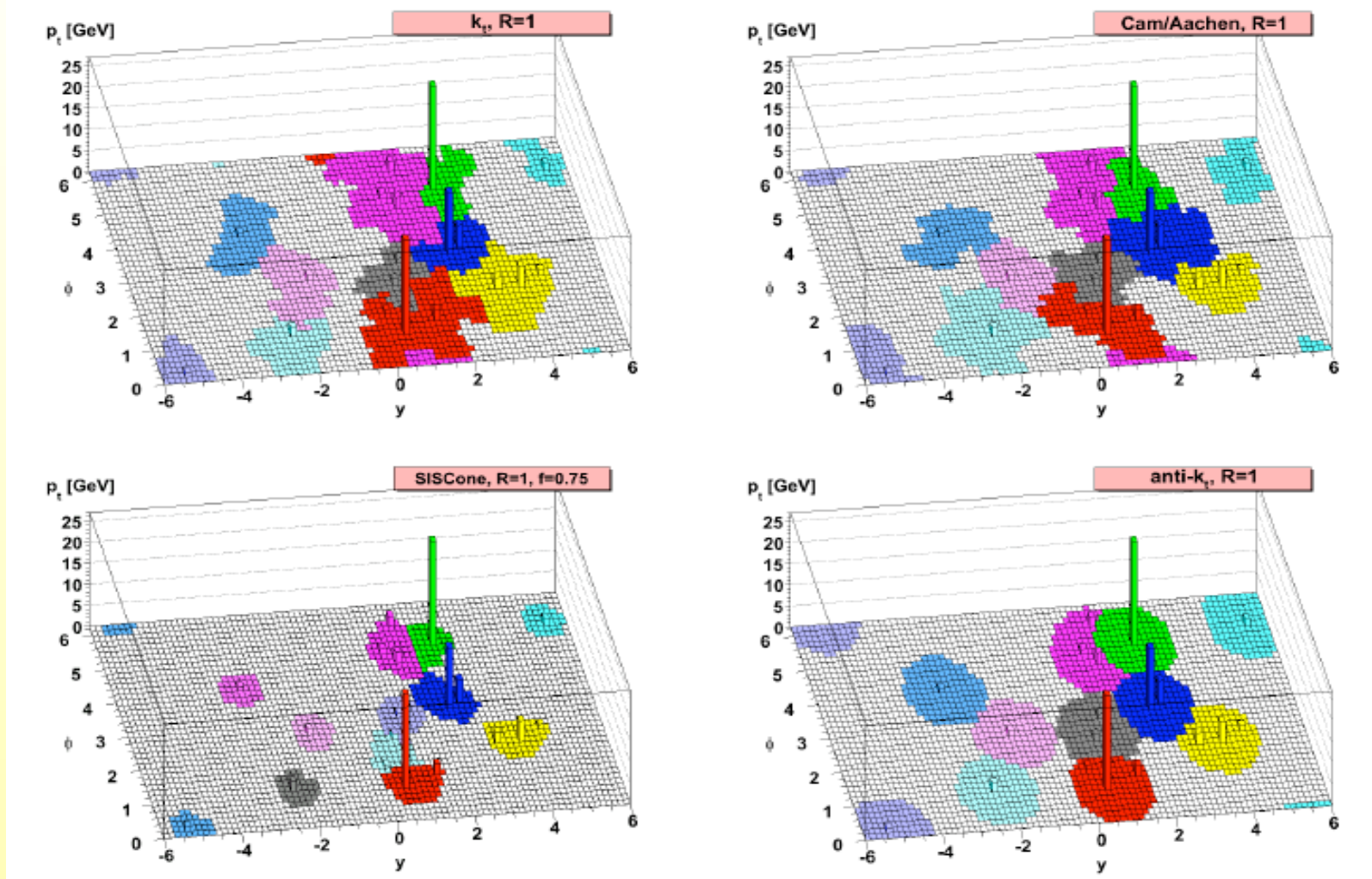


Set of Jet Algorithms

$\text{SR, } d_{ij} = \min(k_{ti}^2, k_{tj}^2) \Delta R_{ij}^2 / R^2$ <p>hierarchical in rel \perp momenta</p>	 <p>Cambridge/Aachen</p> $\text{SR, } d_{ij} = \Delta R_{ij}^2 / R^2$ <p>hierarchical in angle</p>
$\text{SR, } d_{ij} = \min(k_{ti}^{-2}, k_{tj}^{-2}) \Delta R_{ij}^2 / R^2$ <p>gives perfectly conical jets</p>	<p>SISCone</p> <p>Seedless Infrared Safe cone +SM</p> <p>gives "economical" jets</p>

This 3 Algorithms are studied in order to understand their performances under different physics processes and detector's conditions.

Jet Areas



A sample parton-level event, together with many random soft “ghosts”, clustered with four different jets algorithms, illustrating the “active” catchment areas of the resulting hard jets. For K_t and Cam/Aachen the detailed shapes are in part determined by the specific set of ghosts used, and change when the ghosts are modified.

Pileup subtraction using Jet areas

Basic Procedure:

- ▶ Use p_t/A from majority of jets (pileup jets) to get level, ρ , of pileup and UE in event

- ▶ Subtract pileup from hard jets:

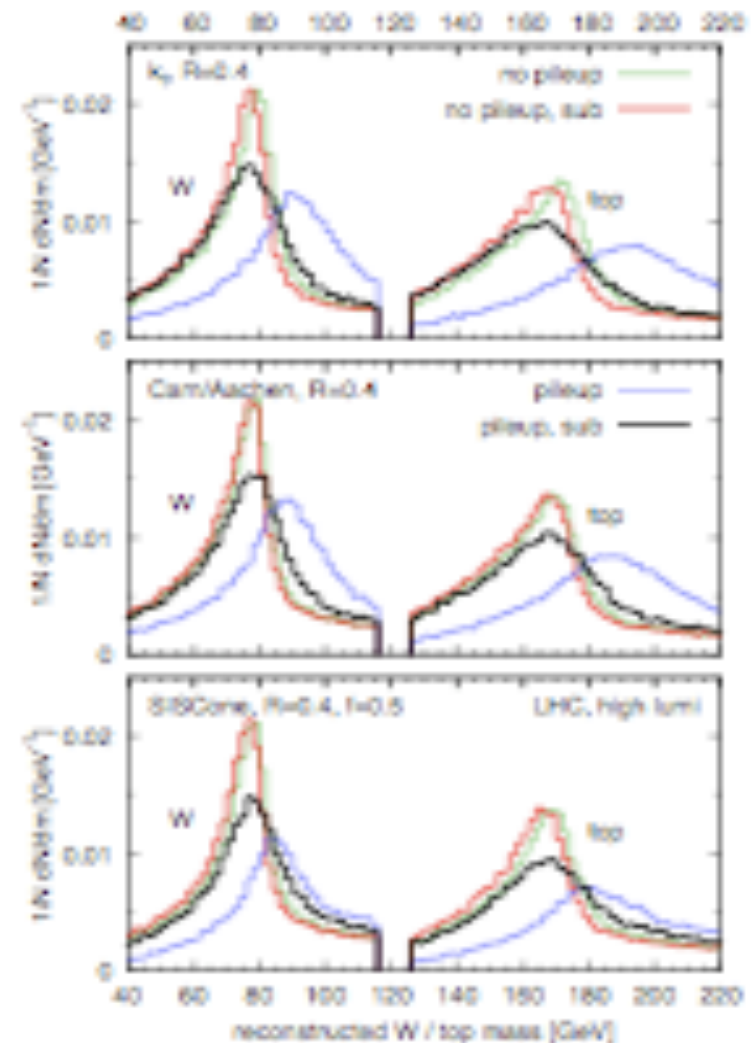
$$p_t \rightarrow p_{t,sub} = p_t - A\rho$$

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Illustration:

- ▶ semi-leptonic $t\bar{t}$ production at LHC
- ▶ high-lumi pileup (~ 20 ev/bunch-X)

Same simple procedure works for a range of algorithms



Invariant mass vs Jet size

$$Q_{w=x\sqrt{M}}^{1/f} \equiv \left(\frac{\text{Max \# reco. massive objects in window of width } w = x\sqrt{M}}{\text{Total \# generated massive objects}} \right)^{-1}$$

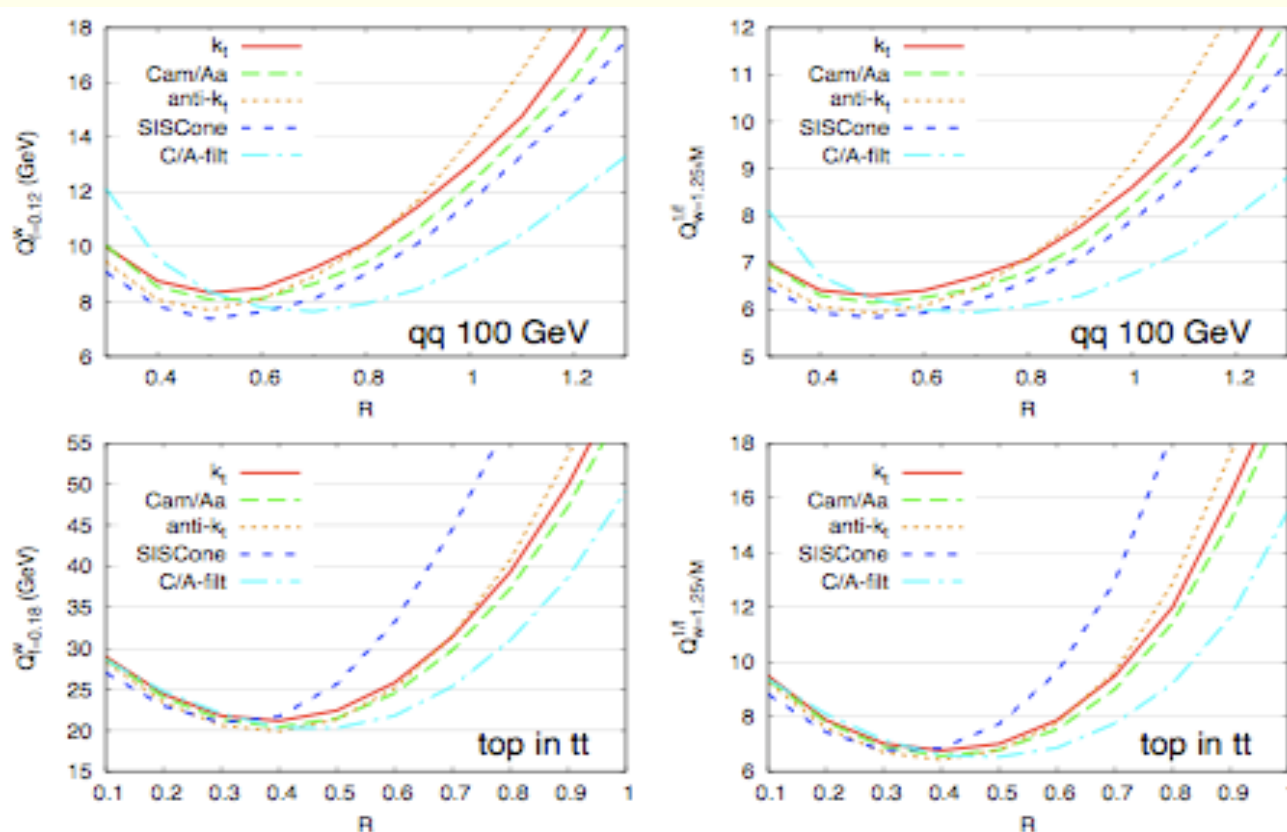


Figure 3: The quality measures $Q_{f=z}^w$ (left) and $Q_{w=1.25\sqrt{M}}^{1/f}$ (right), for different jet algorithms as a function of R , for the 100 GeV $q\bar{q}$ case (top row), 2 TeV $g\bar{g}$ (middle row) and top reconstruction in $t\bar{t}$ events (bottom row).