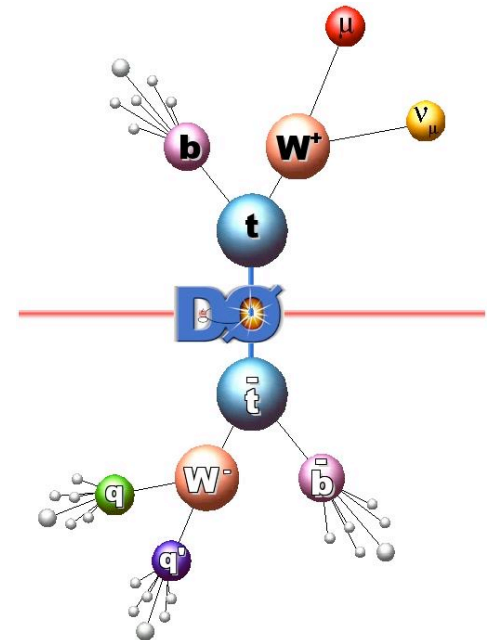
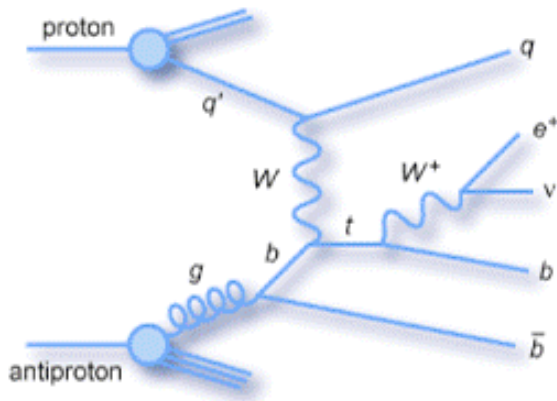


# Top Quark Highlights at the Tevatron

MANCHESTER  
1824



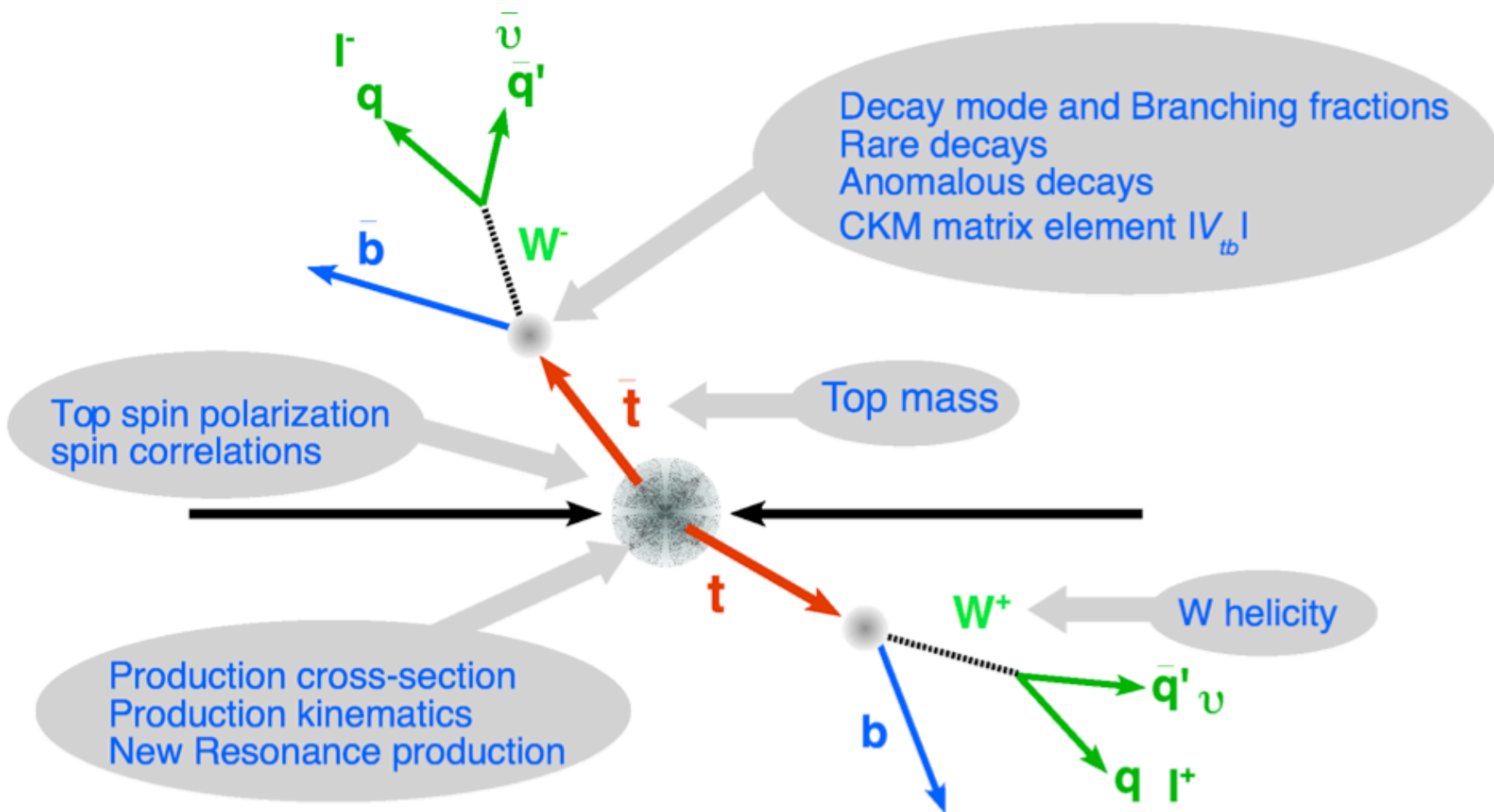
## Christian Schwanenberger



Half day meeting: Tevatron for LHC, 31.01.2007

# Top quark physics

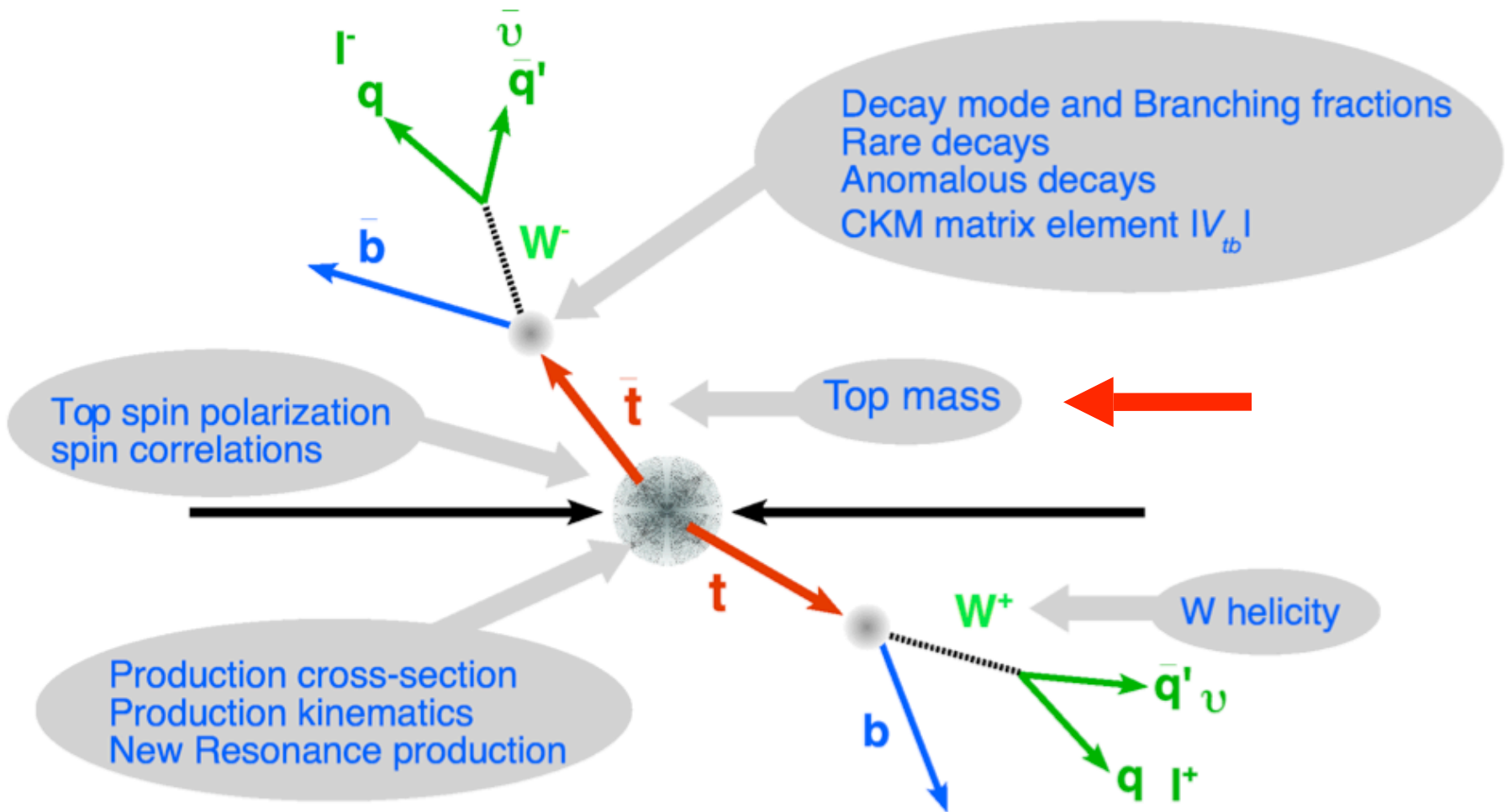
- The Tevatron is still the only place to make top quarks.



- Other predicted production mode: **single top**

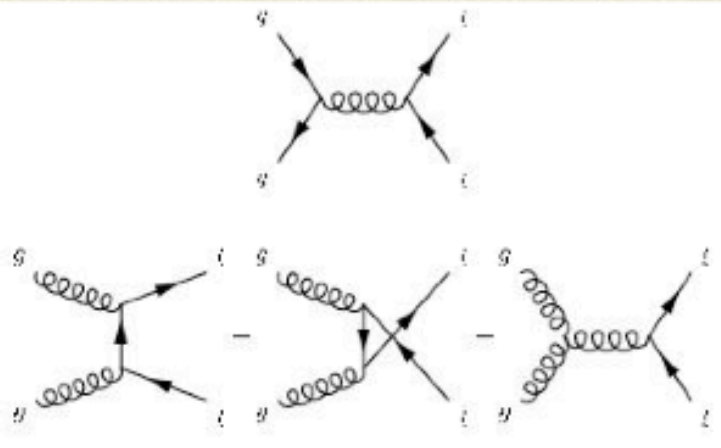
# Top quark physics

- The Tevatron is still the only place to make top quarks.



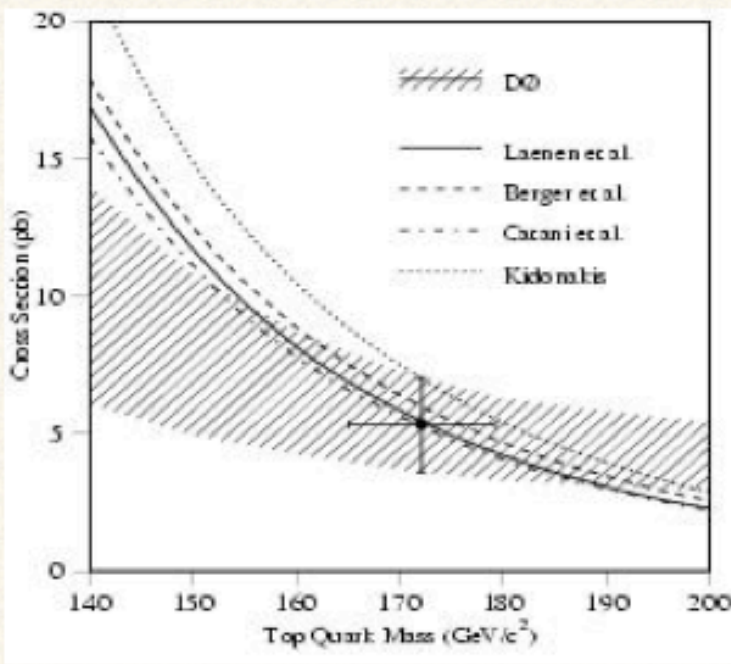
- Other predicted production mode: **single top** ←

# Strong Top Quark Production

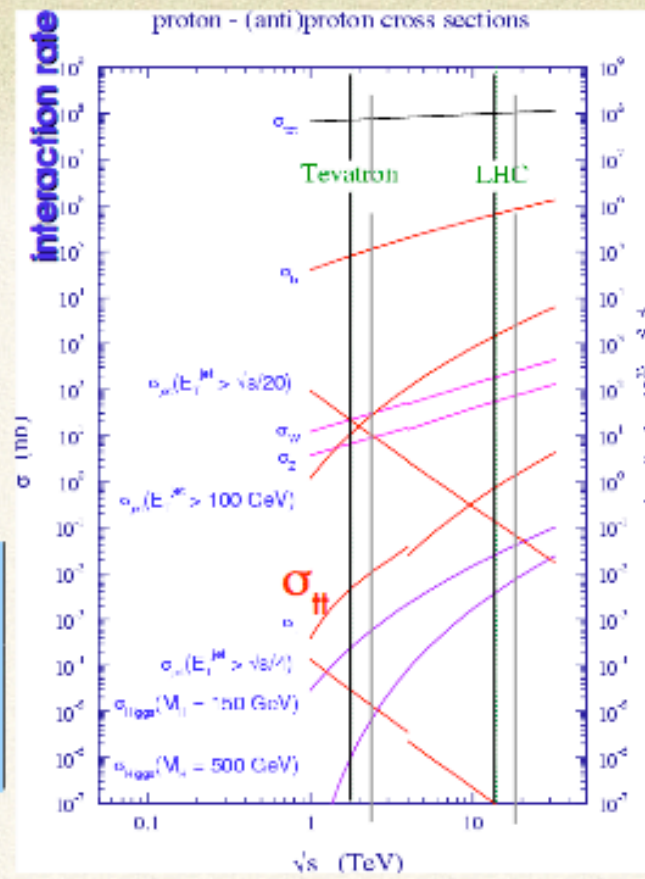


qq ~ 85 %

gg ~ 15 %



	Run I	Run II (2 fb <sup>-1</sup> )	LHC (10 fb <sup>-1</sup> )
no ttbar (m <sub>t</sub> sample ≥ 1 b-tag)	20	800	8 * 10 <sup>6</sup>



- establish top signal
- measure cross section as QCD test
- cross section and topology close to Higgs physics

# Decay Topology in Top-Antitop Production

decay:  $t \rightarrow W b$  (100%)

## Top-Antitop Signatures:

'dilepton channel'

5% : 2 jets, 2 charged leptons, 2  $\nu$

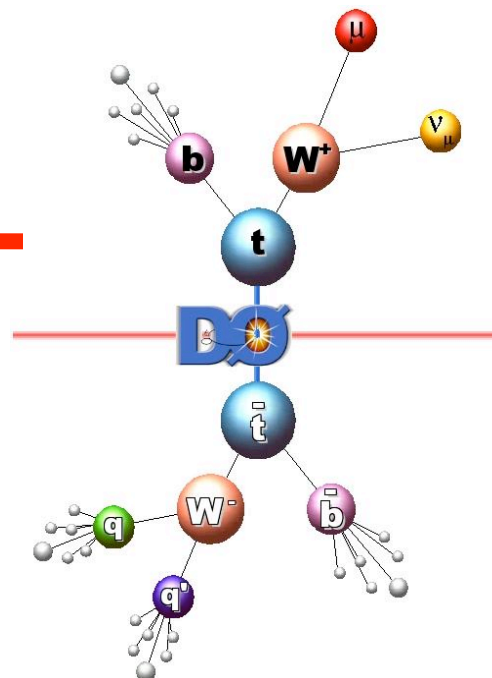
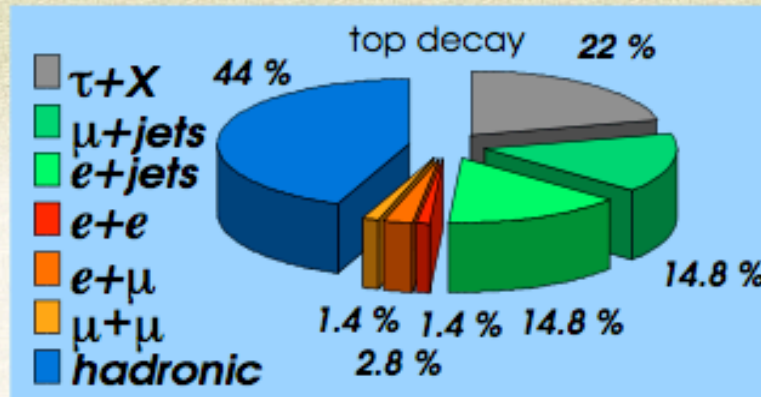
'lepton+jets channel'

30% : 4 jets, 1 charged lepton, 1  $\nu$

'all-jets channel'

40% : 6 jets

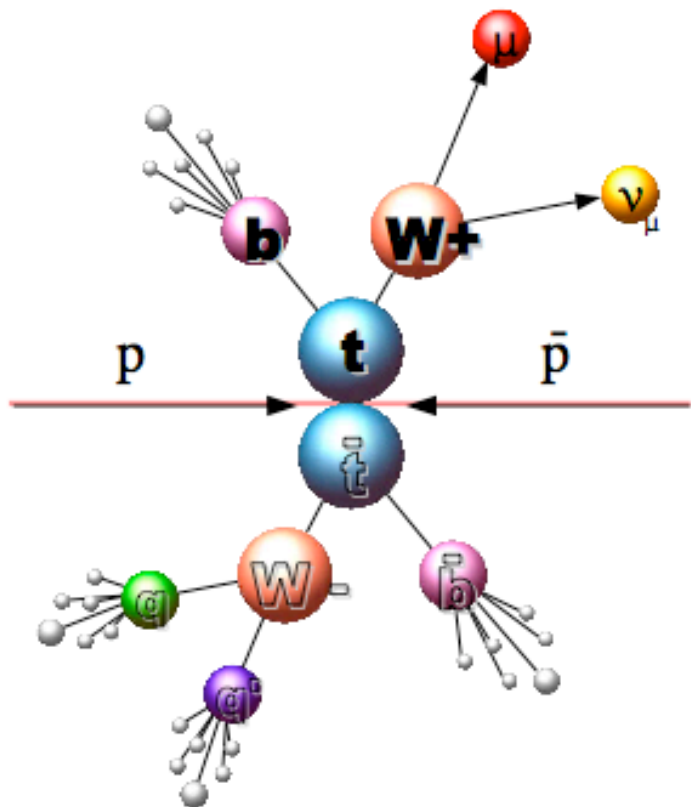
always 2 jets are b-jets



# Event Topology in Lepton+Jets

**signal**

- 1 lepton with high  $p_T$
- 1  $\nu$  (reconstructed as transverse energy = MET)
- $\geq 4$  jets

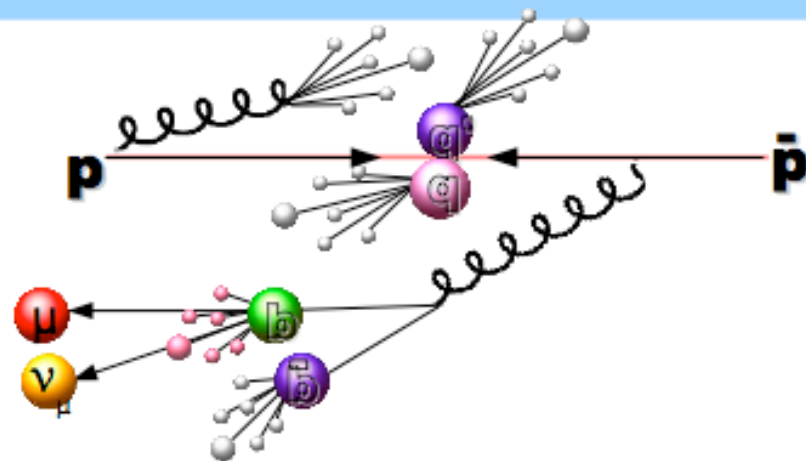


**background**

$W \rightarrow l\nu + \geq 4$  jets

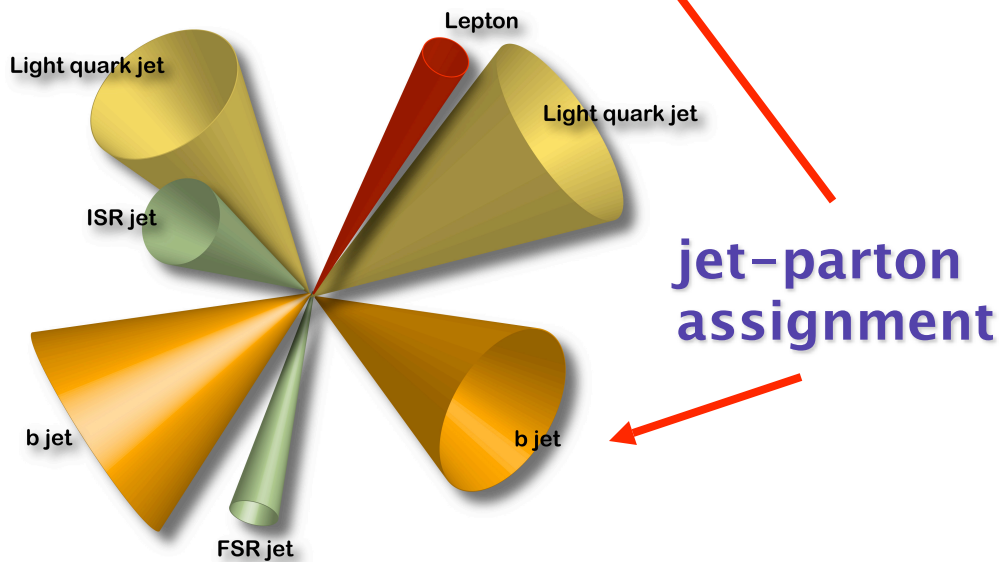
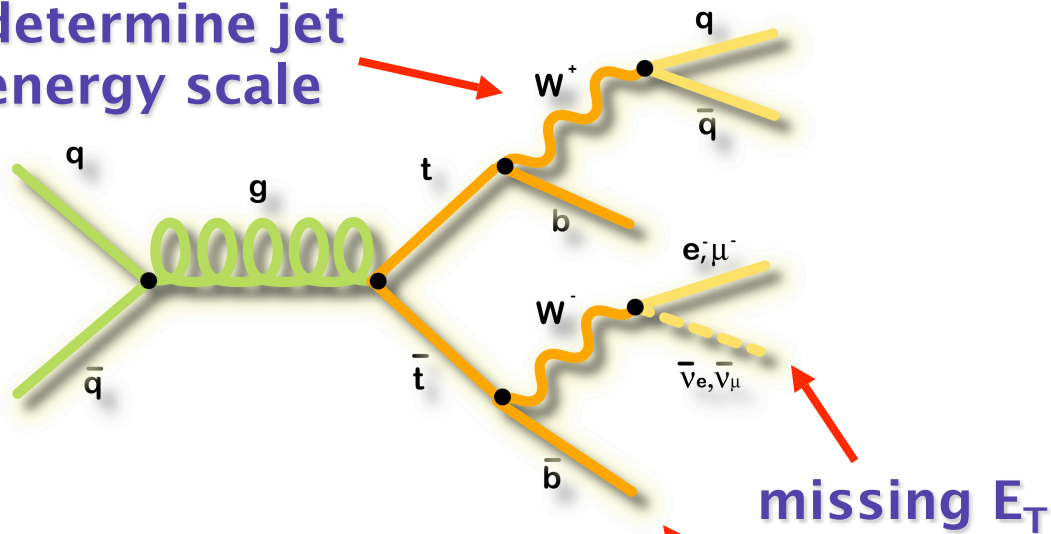


multijet background (QCD)  
+ misreconstructed met  
+ fake isolated  $\mu$  or  $e$

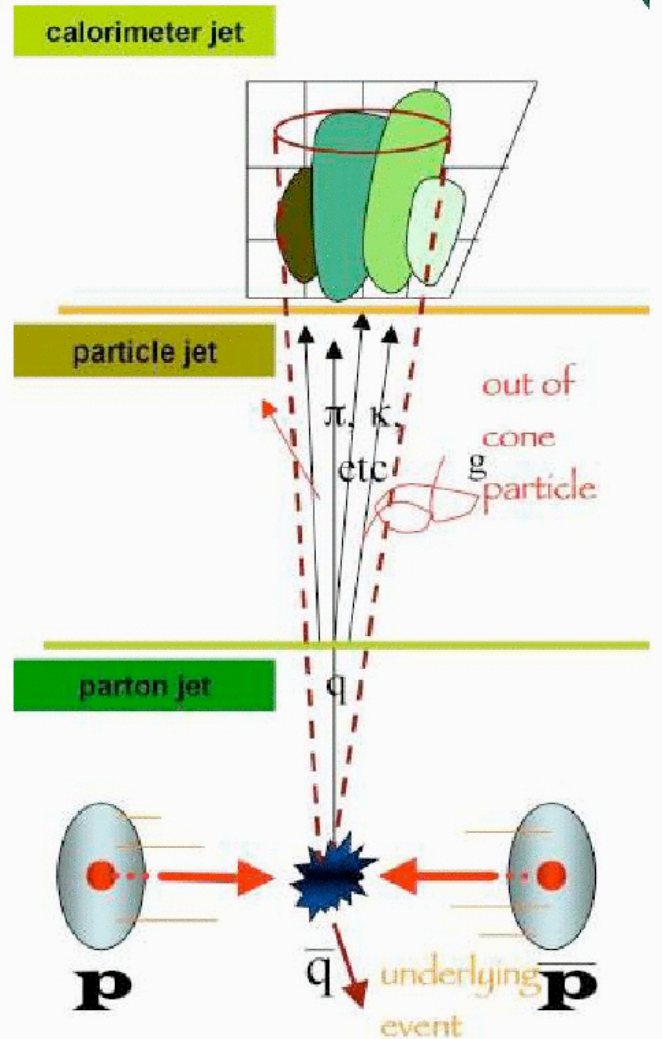


# Lepton+Jets Channel

determine jet energy scale

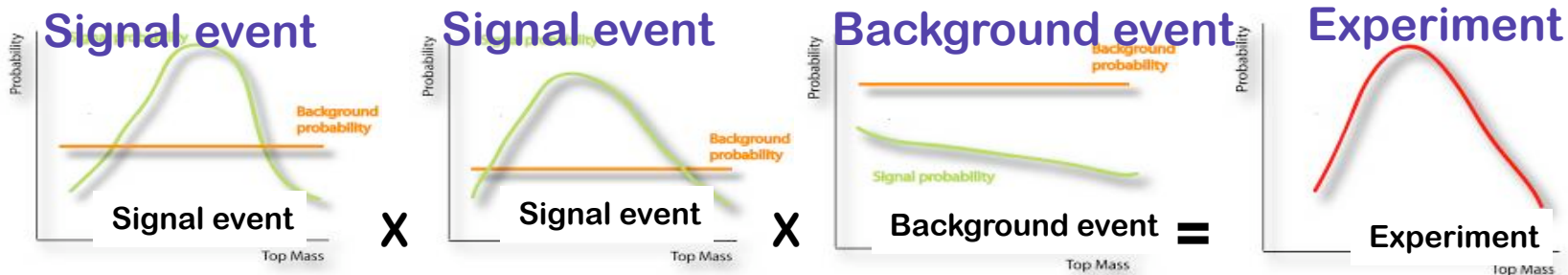
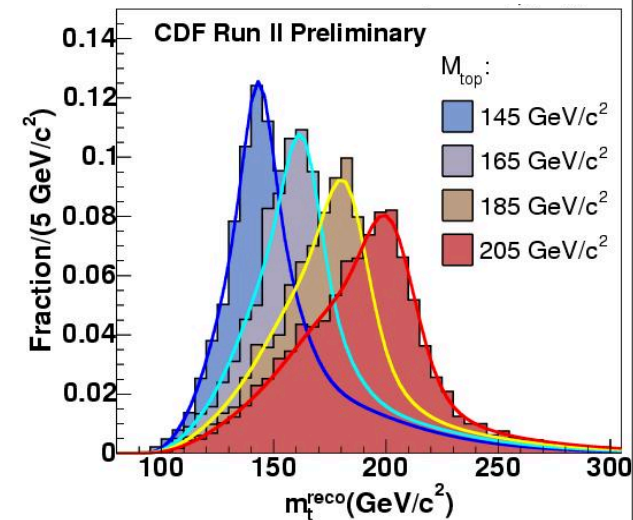


jet energy scale:



# Extraction Techniques

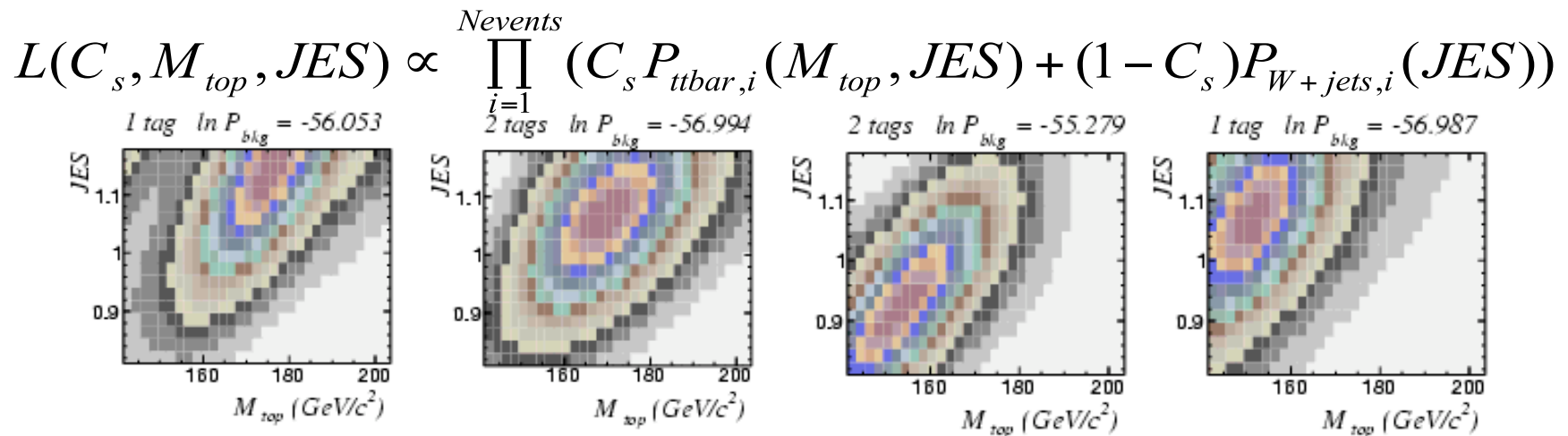
- Need to extract  $M_{\text{top}}$  from imprecise measurements (jets) and non-measured (neutrino) quantities
- **Template analyses:**
  - Evaluate variable strongly correlated with  $M_{\text{top}}$
  - Obtain  $M_{\text{top}}$  comparing data to Monte Carlo with different  $M_{\text{top}}$  input
- **Matrix Element analyses:**
  - Evaluate  $t\bar{t}$  and background probability densities as a function of  $M_{\text{top}}$





# Matrix Element Method

- Most precise measurements by CDF and DØ use the Matrix Element method in the leptons+jets channel with in-situ determination of the jet energy scale
  - Define event likelihood using signal  $P_{t\bar{t}bar}$  and background  $P_{W+jets}$  probability density
  - Use maximum likelihood to fit simultaneously  $M_{top}$ , JES, and signal fraction,  $C_s$



# Matrix Element Method

- CDF has used 940 pb<sup>-1</sup> and measured with 166 candidates with at least one b-tagged jet

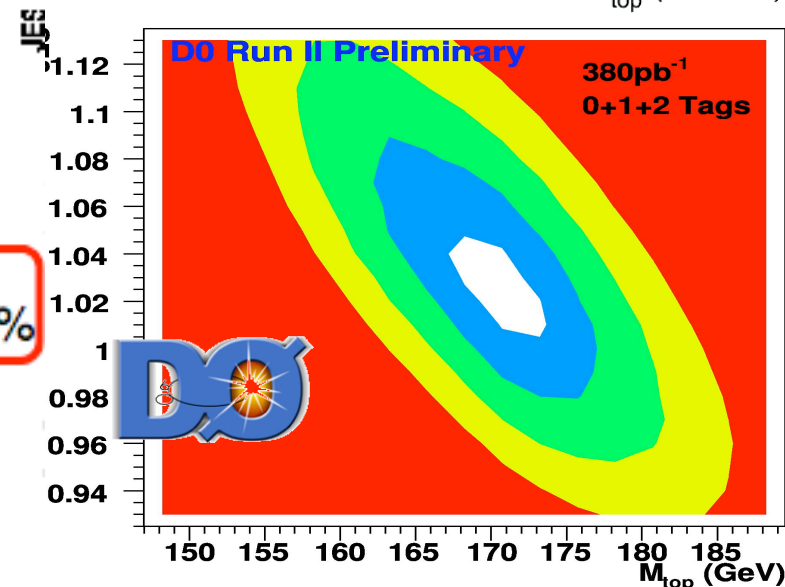
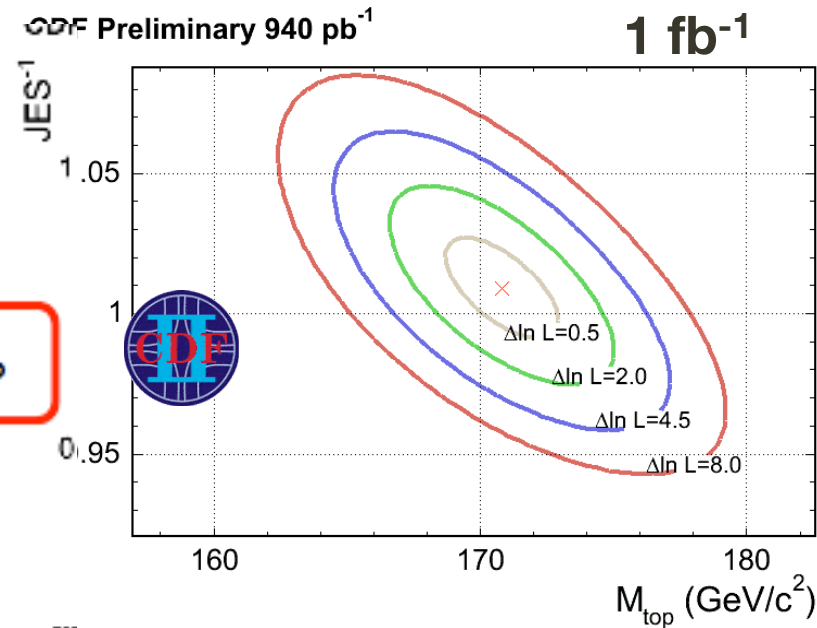
$$M_{top} = 170.9 \pm 1.6 (\text{stat.}) \pm 1.4 (\text{JES}) \pm 1.4 (\text{syst.}) \text{ GeV} / c^2 \quad 1.5\%$$

**World's most precise measurement!**

- DØ has used 380 pb<sup>-1</sup> and measure with 175 candidates with and without b-tagging requirement

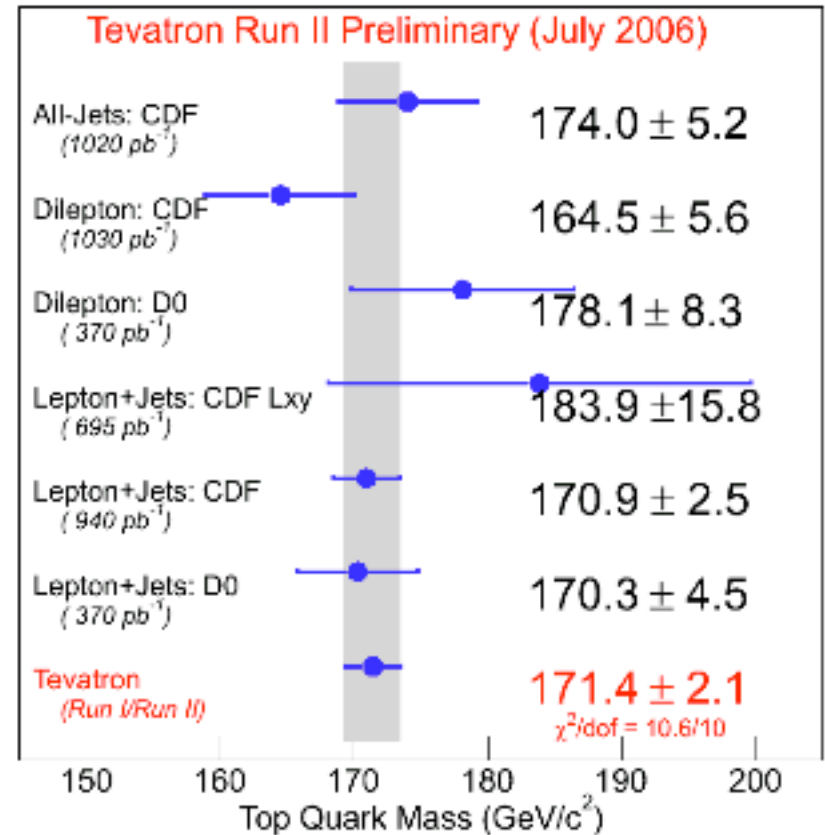
$$M_{top} = 170.3 \pm 2.5 (\text{stat.}) \pm 3.5 (\text{JES}) \pm 1.5 (\text{syst.}) \text{ GeV} / c^2 \quad 2.6\%$$

- Update from DØ coming soon.



# Tevatron Combination

- Excellent results in each channel
- Combine them to improve precision:
  - Include Run-I results
  - Account for correlations
  - Use Best Linear Unbiased Estimator (*NIM A270 110, A500 391*)
- We reached a precision of of 1.2% in  $M_{\text{top}}$



$$M_{\text{top}} = 171.4 \pm 1.2(\text{stat.}) \pm 1.4(\text{JES}) \pm 1.0(\text{syst.}) \text{ GeV}/c^2 \quad 1.2\%$$

# Summary: Top Mass Measurements

- New more precise measurements in every channel from CDF and DØ

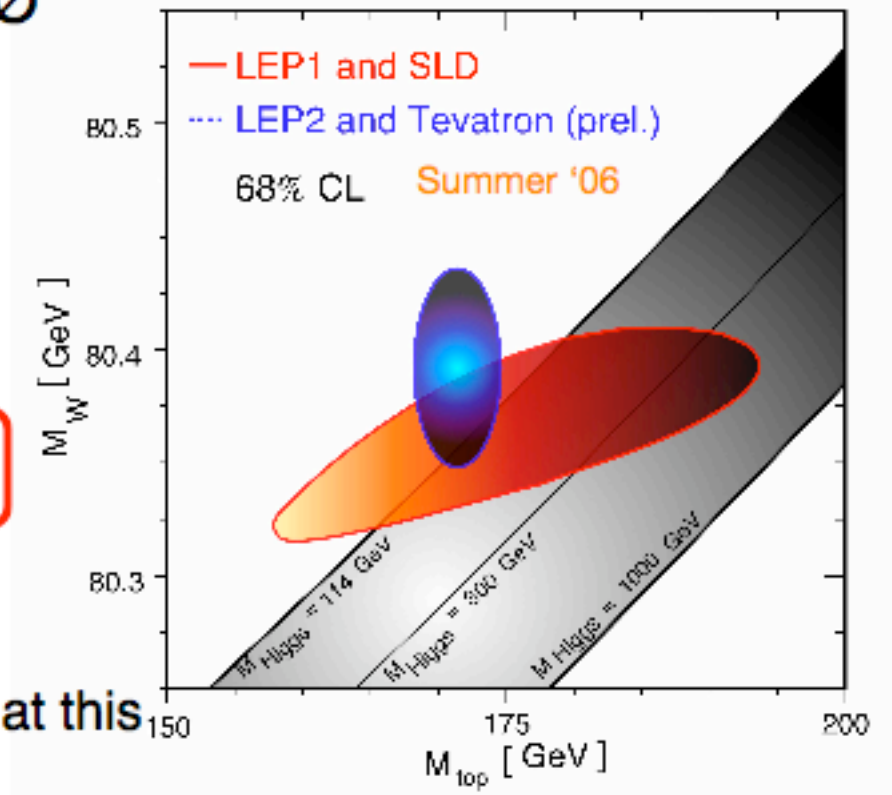
- All-hadronic channel becomes competitive

- New world average

$$M_{\text{top}} = 171.4 \pm 2.1 \text{ GeV}/c^2 \quad 1.2\%$$

- Present uncertainties on  $M_{\text{top}}$  (and  $M_W$ , new results to be presented at this conference!) help constrain  $M_{\text{Higgs}}$  to about  $35\% \delta M_{\text{Higgs}} / M_{\text{Higgs}}$

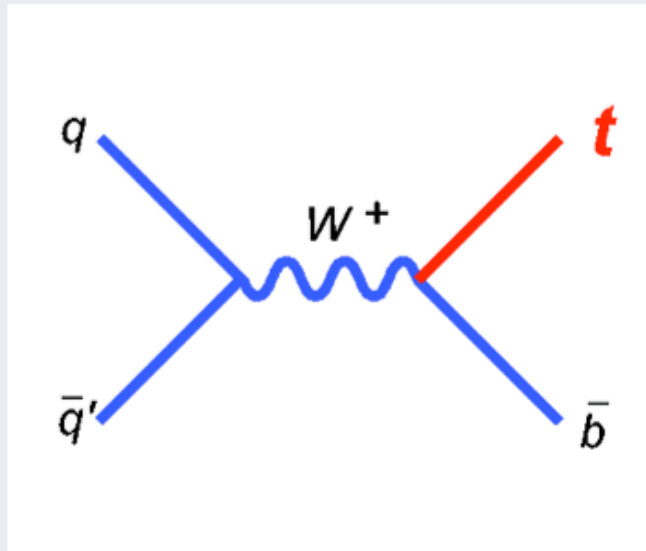
- Tevatron should reach a precision of  $\ll 1\%$  with the full Run II data set



# Single top quark production

- Electroweak production in two main mechanisms at the Tevatron:

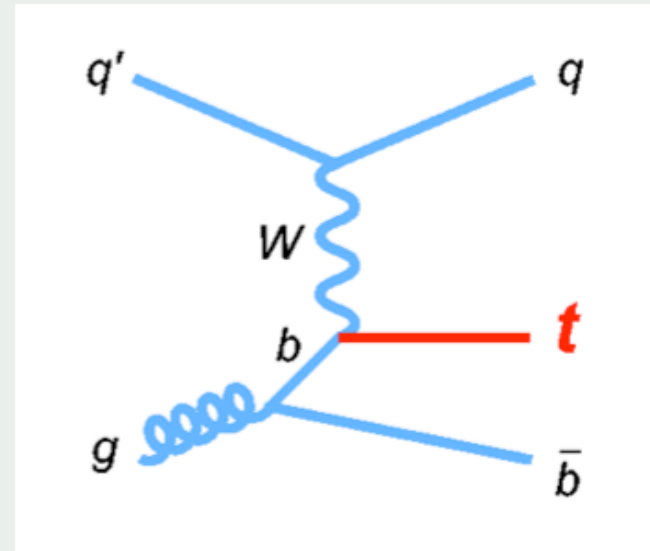
## s-channel (tb)



- $\sigma_{NLO} = 0.88 \pm 0.11 \text{ pb} (*)$
- previous limits (95% C.L.):

Run II DØ:  $< 5.0 \text{ pb} (370 \text{ pb}^{-1})$   
Run II CDF:  $< 3.1 \text{ pb} (700 \text{ pb}^{-1})$

## t-channel (tqb)







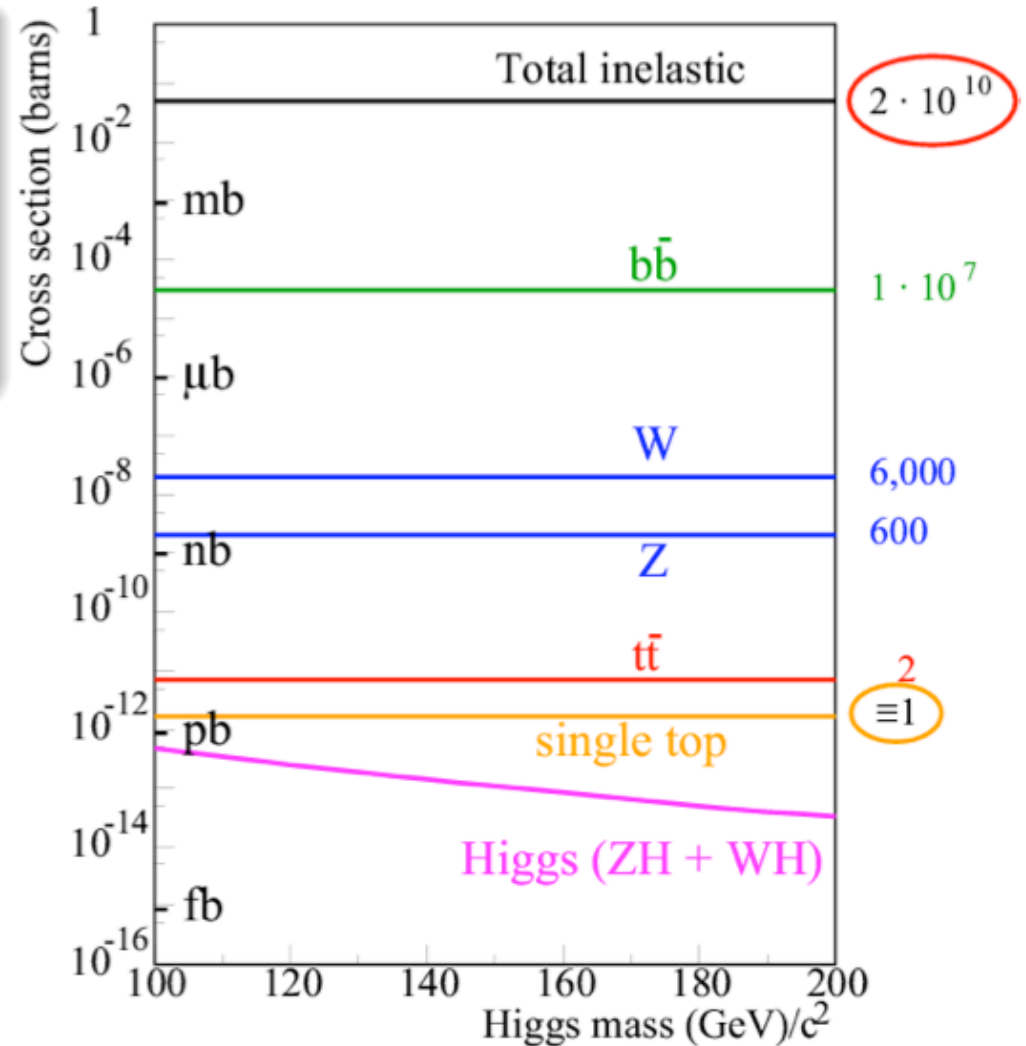
- $\sigma_{NLO} = 1.98 \pm 0.25 \text{ pb} (*)$
- previous limits (95% C.L.):

Run II DØ:  $< 4.4 \text{ pb} (370 \text{ pb}^{-1})$   
Run II CDF:  $< 3.2 \text{ pb} (700 \text{ pb}^{-1})$

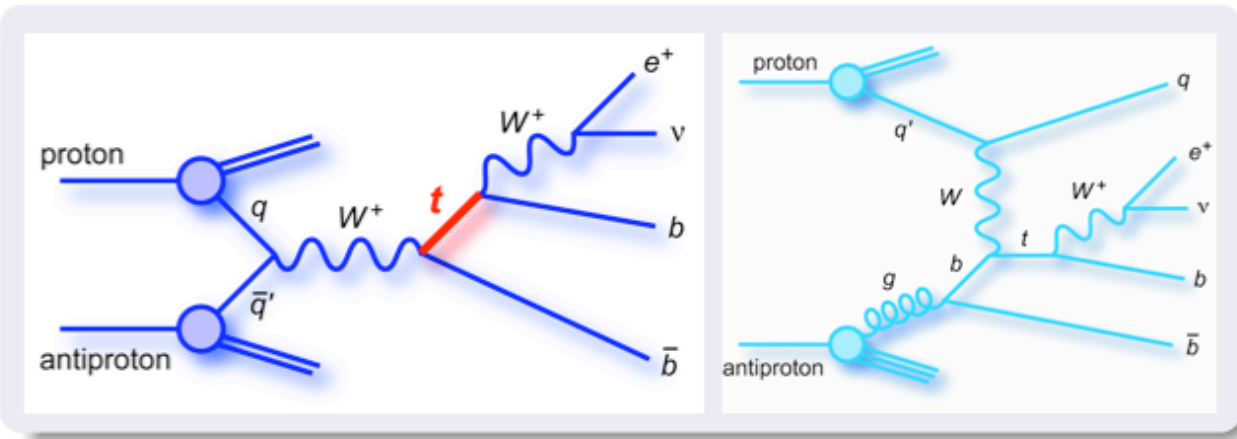
(\*)  $m_t = 175 \text{ GeV}$ , Phys.Rev. D70 (2004) 114012

# It has been challenging for years...

- Several publications since Run I by  and 
- 7  and 6  PhDs
- $\sigma_{t\bar{t}}$  only  $\sim 2 \times \sigma_{\text{single top}}$ , but has striking signature





# Event selection



## Signature

- isolated lepton
- $\cancel{E}_T$
- jets
- at least 1 b-jet

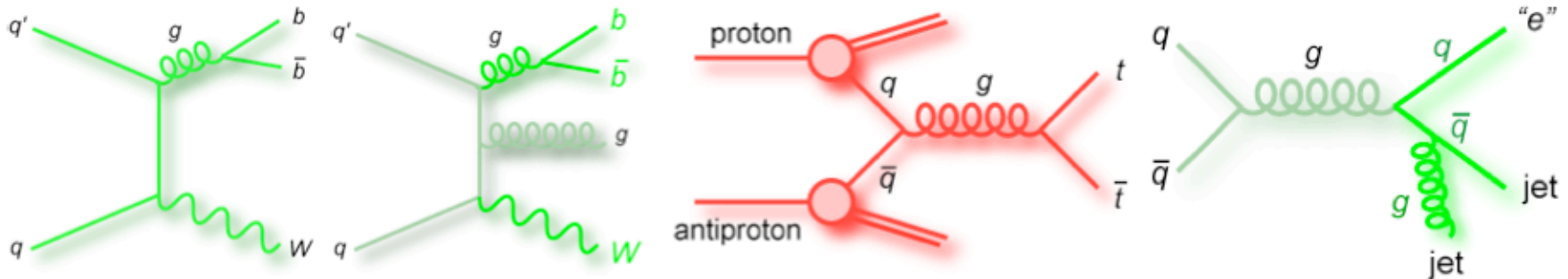
		
1 lepton	$p_T^e > 20 \text{ GeV},  \eta_e  < 2$ $p_T^\mu > 20 \text{ GeV},  \eta_\mu  < 1.1$	$p_T^e > 15 \text{ GeV},  \eta_e  < 1.1$ $p_T^\mu > 18 \text{ GeV},  \eta_\mu  < 2.0$
jets	exactly 2 $p_T > 15 \text{ GeV},  \eta  < 2.8$	2,3,4 $p_T > 15 \text{ GeV},  \eta  < 3.4$ leading jet $p_T > 25 \text{ GeV},  \eta  < 2.5$ 2nd leading jet $p_T > 20 \text{ GeV}$
MET	$\cancel{E}_T > 25 \text{ GeV}$	$15 < \cancel{E}_T < 200 \text{ GeV}$
b jet	one or two	

# Backgrounds

- Slightly different naming conventions and techniques between the two experiments but very similar in the end

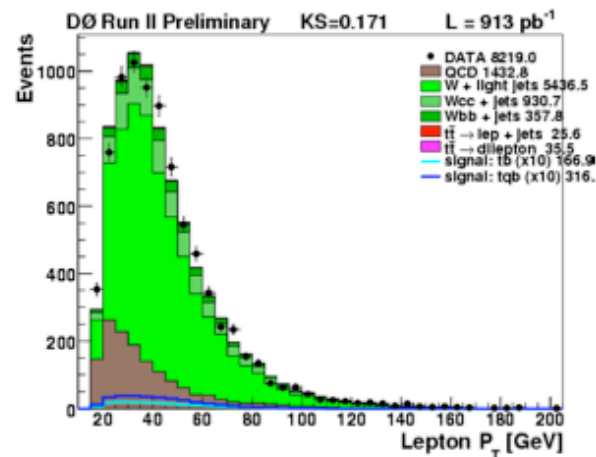
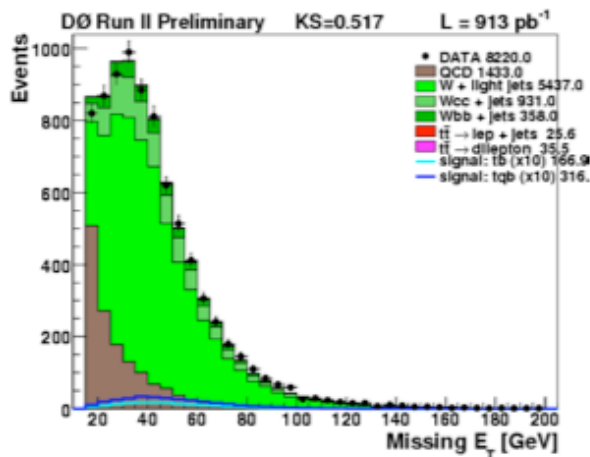
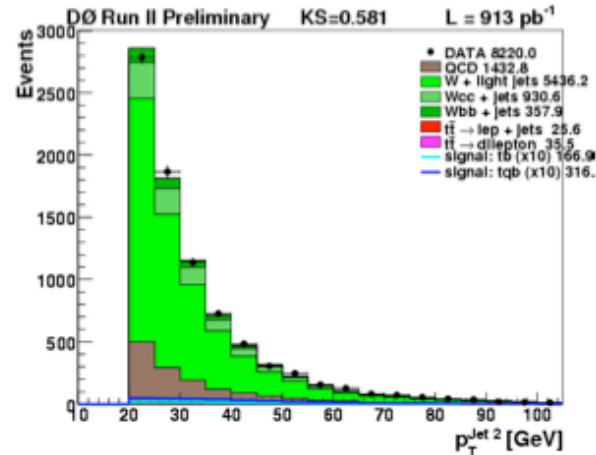
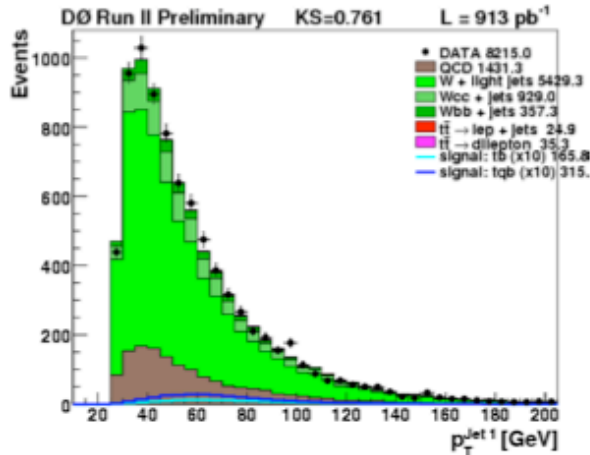
## Main backgrounds

- $W$ +jets (AlpGen, normalized to data):
  - $W$ +heavy flavour:  $Wbb$ ,  $Wbj$ ,  $Wcc$ ,  $Wcj$ ,  $Wc$
  - $W$ +light jets ("mistags")
- $t\bar{t}$  (DO AlpGen, CDF Pythia,  $m_t = 175$  GeV,  $\sigma_{NNLO} = 6.8$  pb)
- QCD (a.k.a. multijet, non- $W$ ) (from data failing lepton ID)





# Event selection – before tagging



- Normalize  $W + \text{multijet}$  to data before tagging
- Checked 90 variables, 3 jet multiplicities, 1-2 tags, electron + muon
- Shown: electron, 2 jets, before tagging
- Good description of data

# Event selection – Yields



Source	Event Yields in 0.9 fb <sup>-1</sup> Data		
	Electron+muon, 1tag+2tags combined		
	2 jets	3 jets	4 jets
<i>tb</i>	16 ± 3	8 ± 2	2 ± 1
<i>tqb</i>	20 ± 4	12 ± 3	4 ± 1
<i>t<math>\bar{t}</math> → ll</i>	39 ± 9	32 ± 7	11 ± 3
<i>t<math>\bar{t}</math> → l+jets</i>	20 ± 5	103 ± 25	143 ± 33
<i>W+b<math>\bar{b}</math></i>	261 ± 55	120 ± 24	35 ± 7
<i>W+c<math>\bar{c}</math></i>	151 ± 31	85 ± 17	23 ± 5
<i>W+jj</i>	119 ± 25	43 ± 9	12 ± 2
Multijets	95 ± 19	77 ± 15	29 ± 6
Total background	686 ± 41	460 ± 39	253 ± 38
Data	697	455	246



<i>s</i> -channel	15.4 ± 2.2
<i>t</i> -channel	22.4 ± 3.6
<i>tt</i>	58.4 ± 13.5
Diboson	13.7 ± 1.9
Z + jets	11.9 ± 4.4
<i>Wbb</i>	170.9 ± 50.7
<i>Wcc</i>	63.5 ± 19.9
<i>Wc</i>	68.6 ± 19.0
Non- <i>W</i>	26.2 ± 15.9
Mistags	136.1 ± 19.7
Single top	37.8 ± 5.9
Total background	549.3 ± 95.2
Total prediction	587.1 ± 96.6
Observed	644

- Expected single top signal is smaller than background uncertainty!  
⇒ No counting experiment, requires advanced analysis techniques

# Multivariate analysis techniques

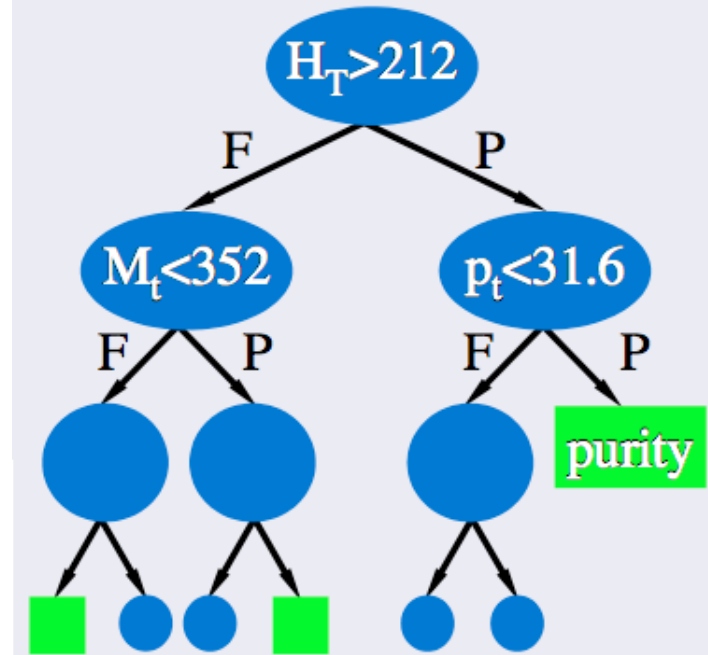
- Likelihood discriminants (🇮🇹)

- Artificial neural network (🇮🇹)

- Matrix element (🇩🇪, 🇮🇹)

- Bayesian neural networks (🇩🇪)

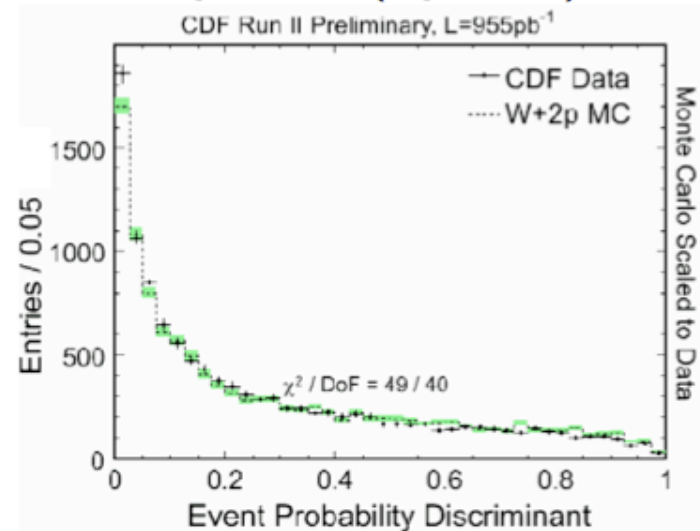
- Boosted decision trees (🇩🇪)



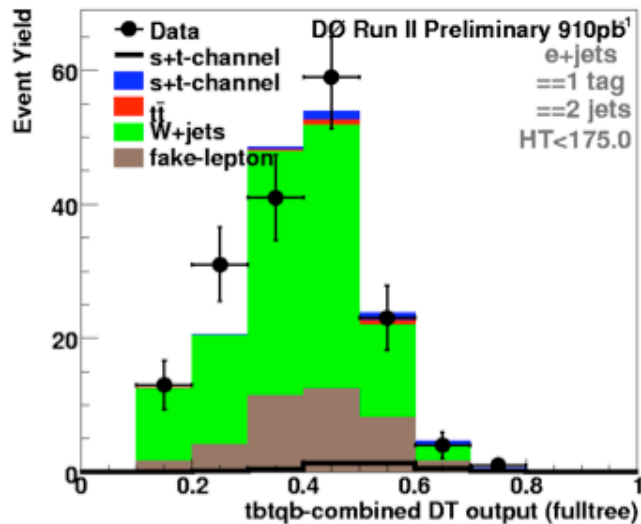
# Cross-check samples

- Validate methods using data without looking at signal
- Compare discriminant in model and data
- Good agreement observed

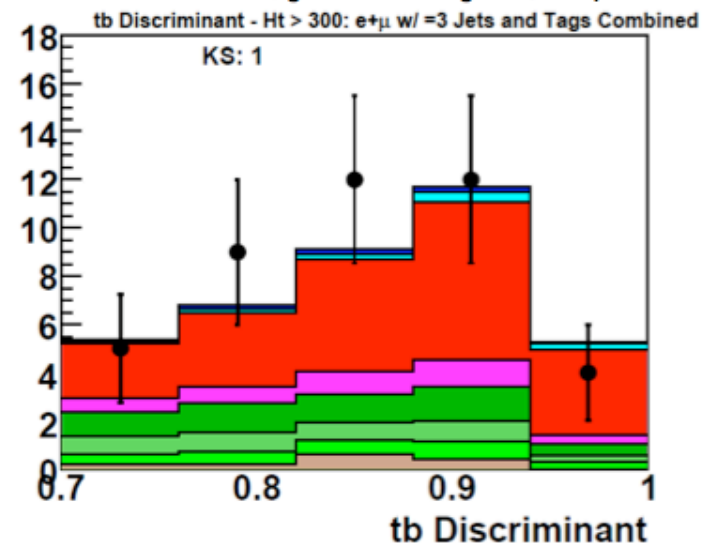
ME  $W+2$ jets data ( $b$ -jet veto)



DT " $W$ +jets": =2jets,  $H_T < 175$  GeV

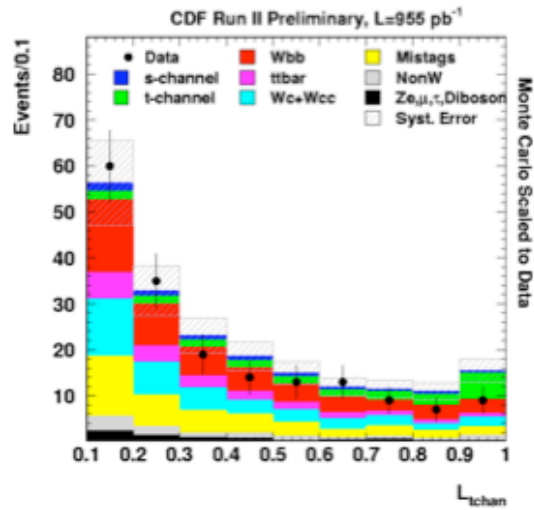


ME " $W$ +jets": =3jets,  $H_T > 300$  GeV



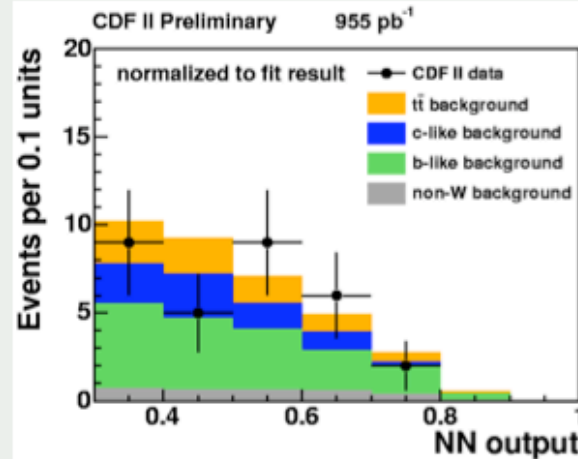
# CDF s+t observed results – Preliminary

## Likelihood



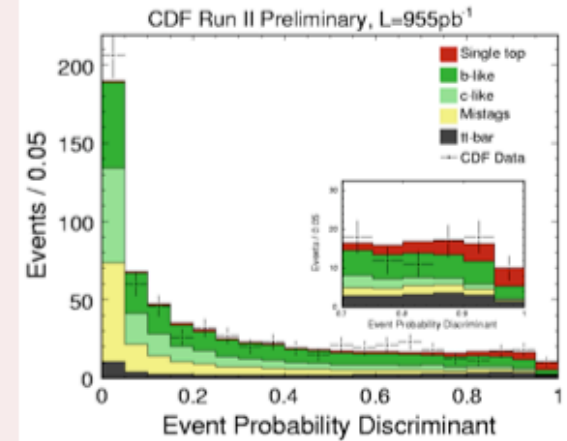
No evidence of signal  
 $\sigma < 2.7 \text{ pb @ 95\% CL}$   
 From s and t likelihoods

## Neural network



no evidence of signal  
 $\sigma < 2.6 \text{ pb @ 95\% CL}$

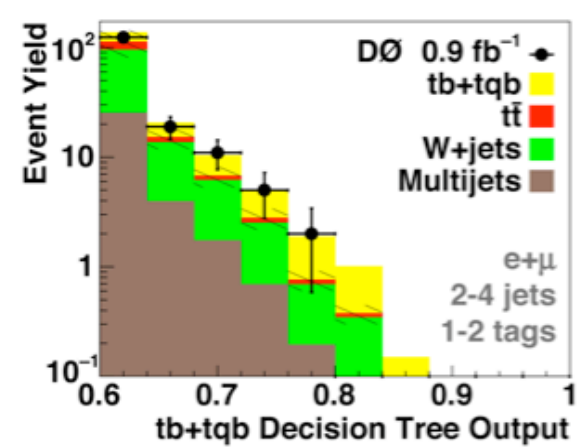
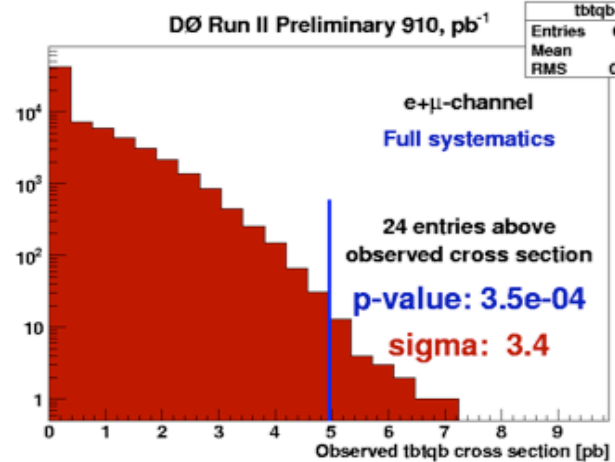
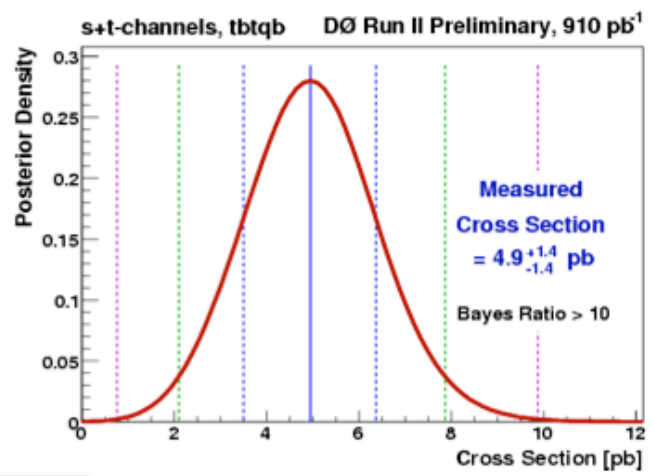
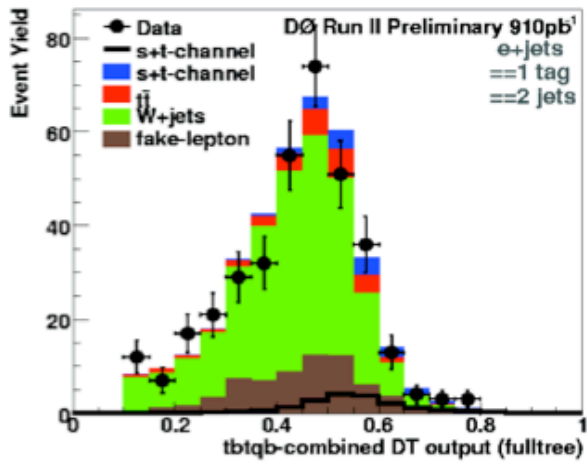
## Matrix element



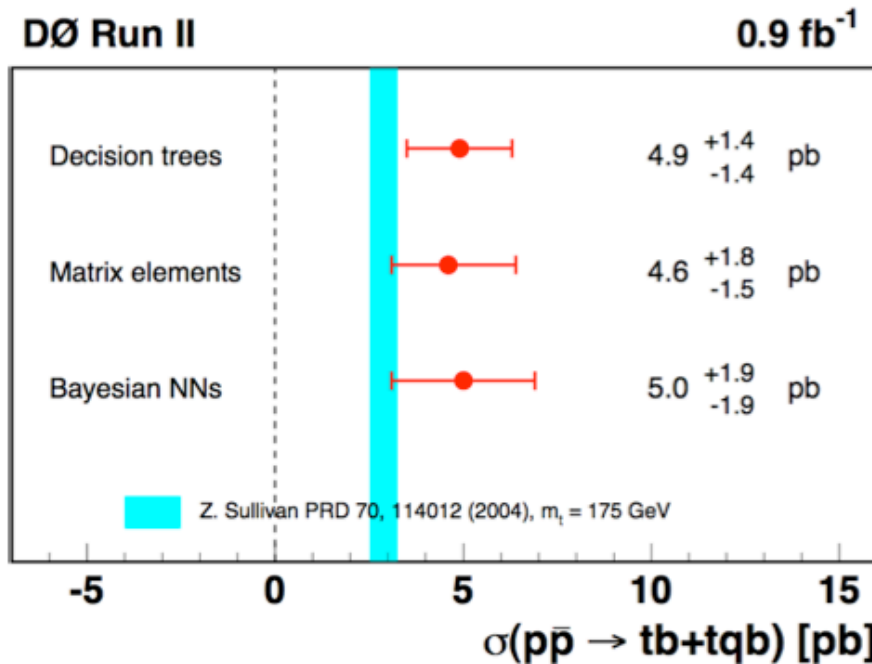
$\sigma = 2.7^{+1.5}_{-1.3} \text{ pb}$   
 $p\text{-value} = 1.0\% (2.3\sigma)$

# DØ boosted decision tree s+t observed results

$\sigma = 4.9 \pm 1.4 \text{ pb}$     expected:  $2.9^{+1.6}_{-1.4} \text{ pb}$   
 $p\text{-value} = 0.035\% (3.4\sigma)$   
 SM compatibility: 11% ( $1.1\sigma$ )



# DØ results consistency



## High discriminant correlation

Choose the 50 highest events in each discriminant and look for overlap

	Electron	Muon
DT vs ME	52%	58%
DT vs BNN	56%	48%
ME vs BNN	46%	52%

## Linear correlation

Measured cross section in 400 members of SM ensemble with all three techniques and calculated the linear correlation between each pair

	DT	ME	BNN
DT	100%	39%	57%
ME		100%	29%
BNN			100%

# First direct measurements of $|V_{tb}|$

## Direct access to $|V_{tb}|$

$$V_{CKM} = \begin{pmatrix} V_{ud} & V_{us} & V_{ub} \\ V_{cd} & V_{cs} & V_{cb} \\ V_{td} & V_{ts} & V_{tb} \end{pmatrix}$$

- Weak interaction eigenstates are not mass eigenstates
- In SM: top must decay to a  $W$  and  $d$ ,  $s$  or  $b$  quark
  - $V_{td}^2 + V_{ts}^2 + V_{tb}^2 = 1$
  - constraints on  $V_{td}$  and  $V_{ts}$ :  $|V_{tb}| = 0.9991$
- New physics:
  - $V_{td}^2 + V_{ts}^2 + V_{tb}^2 < 1$
  - no constraint on  $V_{tb}$

## Result

- Translate  $tb+tb$  cross section into measurement of the strength of  $V-A$  coupling  $|V_{tb}f_1^L|$  in  $Wtb$  vertex ( $f_1^L$ : arbitrary left-handed form factor)
- Assume  $V_{td}^2 + V_{ts}^2 \ll V_{tb}^2$  and pure  $V-A$  and CP-conserving  $Wtb$  interaction

$$|V_{tb}f_1^L| = 1.3 \pm 0.2$$

- Also assuming  $f_1^L = 1$ :

$$0.68 < |V_{tb}| \leq 1 \text{ @ 95\% CL}$$

- No assumption about number of quark families or CKM matrix unitarity



# Conclusions

- CDF and DØ have been searching for single top signal for years
- A lot of energy invested in the experimental challenges
  - very small signal hidden in enormous background
  - efficient  $b$ -tagging
  - background modeling (involving data and Monte Carlo)
- Several multivariate techniques being used
- CDF analyses have good sensitivity but got unlucky ( $2.3\sigma$  signal with ME, LF and NN don't see any single top)
- DØ BNN and ME analyses see  $2.4\sigma$  and  $2.9\sigma$  signal

# Summary

## First evidence for single top quark production (DØ decision trees)

$$\sigma(p\bar{p} \rightarrow tb + X, tqb + X) = 4.9 \pm 1.4 \text{ pb}$$

3.4 $\sigma$  significance

## First direct measurement of $|V_{tb}|$ (DØ decision trees)

$$|V_{tb}f_1^L| = 1.3 \pm 0.2$$

assuming  $f_1^L = 1$ :  $0.68 < |V_{tb}| \leq 1$  @ 95% CL

(Always assuming  $V_{td}^2 + V_{ts}^2 \ll V_{tb}^2$  and pure  $V-A$  and CP-conserving  $Wtb$  interaction)

hep-ex/0612052, submitted to PRL

- Working on understanding correlations and on combinations
- A lot more data already at hand

# Outlook: Top Physics at the LHC

quantity	CDF/DØ	ATLAS/CMS
$\Delta\sigma_{t\bar{t}}/\sigma_{t\bar{t}}$	11% with 1 fb <sup>-1</sup> [554]	5%–10% luminosity systematics dominated
$\Delta\sigma_{\text{single-top}}/\sigma_{\text{single-top}}$	26% with 1 fb <sup>-1</sup> [554]	10% (< 2% stat. error with 10 fb <sup>-1</sup> ) ←
$B(t \rightarrow Wb)$	3.3% with 1 fb <sup>-1</sup> [554]	
$V_{tb}$ from $\sigma_{\text{single-top}}$	14% with 1 fb <sup>-1</sup> [554]	6.5%
$V_{tb}$ from $B(t \rightarrow Wb)$	> 0.22 with 1 fb <sup>-1</sup> [554]	0.2% (stat. only)
single-top polarisation	–	1.6% with 10 fb <sup>-1</sup>
$\Delta m_{\text{top}}/m_{\text{top}}$	≤ 2 GeV/c <sup>2</sup> Sect. 7	≈ 1 GeV/c <sup>2</sup> ←
spin correlation $\theta$	40% (2 fb <sup>-1</sup> ) [538]	7% ( $ll \oplus l + \text{jets}$ ) for 10 fb <sup>-1</sup>
spin correlation $\phi$	–	4% ( $ll \oplus l + \text{jets}$ ) for 10 fb <sup>-1</sup>
W-helicity $\mathcal{F}_0$	6.5% with 1 fb <sup>-1</sup> [554]	2%–5% with 10 fb <sup>-1</sup>
W-helicity $\mathcal{F}_+$	2.6% with 1 fb <sup>-1</sup> [554]	1% with 10 fb <sup>-1</sup>
electric charge $q_t$	distinguish $\frac{2}{3}$ and $\frac{4}{3}$ cases with 1 fb <sup>-1</sup> Sect. 7.2	distinguish $\frac{2}{3}$ and $\frac{4}{3}$ cases with 10 fb <sup>-1</sup>
Yukawa coupling $y_t$	–	4.8σ, 16% (12%) with 30(100) fb <sup>-1</sup> ←
FCNC $B(t \rightarrow gq)$	< 1.9 × 10 <sup>-2</sup> with 2 fb <sup>-1</sup> [288, 555]	< 1 × 10 <sup>-5</sup> - < 1.4 × 10 <sup>-3</sup> (10 fb <sup>-1</sup> )
FCNC $B(t \rightarrow Zq)$	< 1.5 × 10 <sup>-2</sup> with 1 fb <sup>-1</sup> [554]	< 6.5 × 10 <sup>-4</sup> - 1.3 × 10 <sup>-3</sup> with 10 fb <sup>-1</sup>
FCNC $B(t \rightarrow \gamma q)$	< 3.0 × 10 <sup>-3</sup> with 1 fb <sup>-1</sup> [554]	< 8.6 × 10 <sup>-5</sup> - 1.9 × 10 <sup>-4</sup> with 10 fb <sup>-1</sup>
FCNC $B(t \rightarrow WbZ)$	–	< 10 <sup>-7</sup> with 100 fb <sup>-1</sup>
$\Delta\sigma^{M_{Z'}=1 \text{ TeV}/c^2}$	100 fb with 1 fb <sup>-1</sup> [554]	700 fb with 30 fb <sup>-1</sup>
$B(Z' \rightarrow t\bar{t})$		
anom. coupling	$F_{2L} \begin{matrix} >+0.55 \\ <-0.18 \end{matrix}$ [553]	$F_{2L} \begin{matrix} >+0.097 \\ <-0.052 \end{matrix}$
	$F_{2R} \begin{matrix} >+0.25 \\ <-0.24 \end{matrix}$ [553]	$F_{2R} \begin{matrix} >+0.13 \\ <-0.12 \end{matrix}$
$\Delta F_{1V,A}^Z$	– [542]	15%–85% (300 fb <sup>-1</sup> )
$\Delta F_{1V,A}^\gamma$	<+1.03...+2.60 >-1.17...-1.88 (8 fb <sup>-1</sup> ) [542]	15%–50% (30 fb <sup>-1</sup> ), 4%–7% (300 fb <sup>-1</sup> )
$\Delta F_{2V,A}^\gamma$	– [542]	35% (30 fb <sup>-1</sup> ), 20% (300 fb <sup>-1</sup> )
$\Delta F_{2V,A}^Z$	– [542]	55% (300 fb <sup>-1</sup> )

A. Quadt  
Eur. Phys.  
J. C48, 835  
(2006)

# Outlook: Top Physics at the LHC

quantity	CDF/DØ	ATLAS/CMS
$\Delta\sigma_{t\bar{t}}/\sigma_{t\bar{t}}$	11% with $1 \text{ fb}^{-1}$	[554] 5%–10% luminosity systematics dominated
$\Delta\sigma_{\text{single-top}}/\sigma_{\text{single-top}}$	26% with $1 \text{ fb}^{-1}$	[554] 10% (< 2% stat. error with $10 \text{ fb}^{-1}$ )
$B(t \rightarrow W)$		
$V_{tb}$ from ...		
$V_{tb}$ from ...		
single-top		
$\Delta m_{\text{top}}/m_{\text{top}}$		
spin corre		
spin corre		
W-helicit		
W-helicit		
electric ch		
Yukawa c		
FCNC B		
FCNC B		
FCNC B		
FCNC B		
$\Delta\sigma^{M_{Z'}=}$		
$B(Z' \rightarrow t)$		
anom. co		
	$F_{2R}^{>+0.25}$ $F_{2R}^{<-0.24}$	[553] $F_{2R}^{>+0.13}$ $F_{2R}^{<-0.12}$
$\Delta F_{1V,A}^Z$	-	[542] 15%–85% ( $300 \text{ fb}^{-1}$ )
$\Delta F_{1V,A}^\gamma$	$<+1.03\dots+2.60$ $>-1.17\dots-1.88$ ( $8 \text{ fb}^{-1}$ )	[542] 15%–50% ( $30 \text{ fb}^{-1}$ ), 4%–7% ( $300 \text{ fb}^{-1}$ )
$\Delta F_{2V,A}^\gamma$	-	[542] 35% ( $30 \text{ fb}^{-1}$ ), 20% ( $300 \text{ fb}^{-1}$ )
$\Delta F_{2V,A}^Z$	-	[542] 55% ( $300 \text{ fb}^{-1}$ )

in addition:

- detector commissioning studies
- tests of higher order QCD
- b-fragmentation
- ...

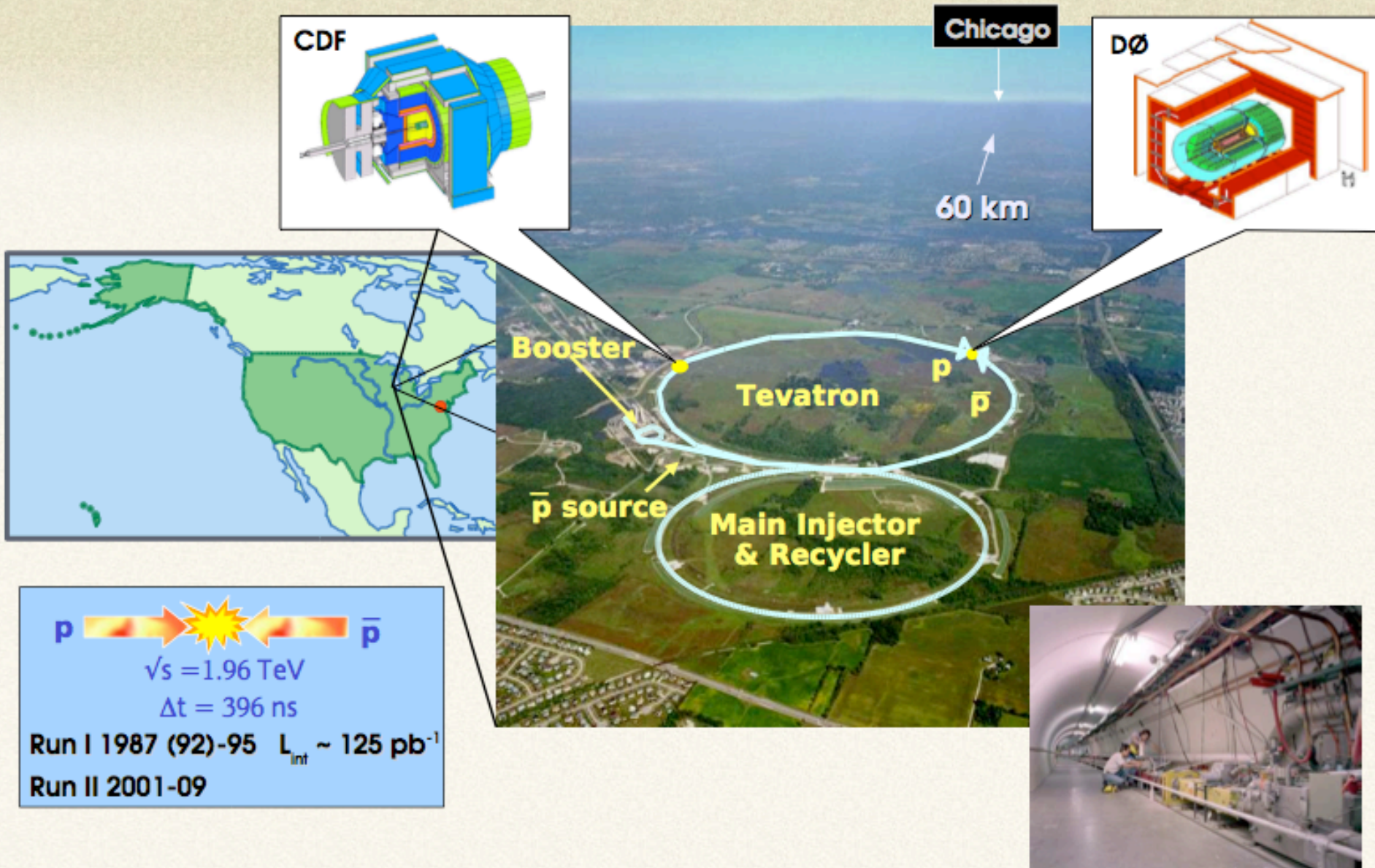
⇒ very interesting top physics at LHC

at  
Early Phys.  
J. C48, 835  
(2006)

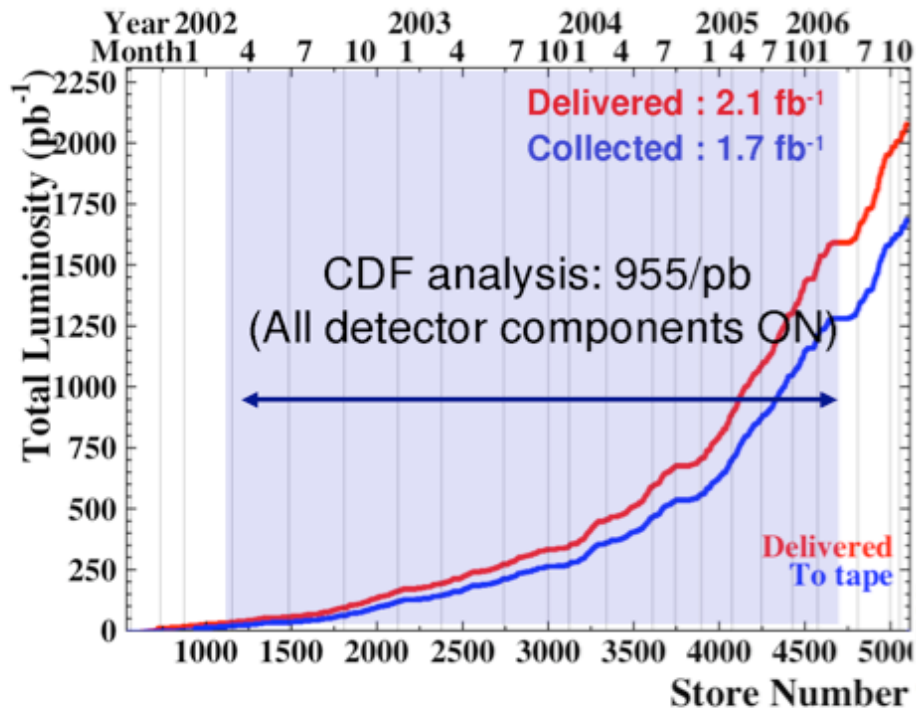


# Backup

# The Tevatron at Fermilab: $p\bar{p}$ collisions

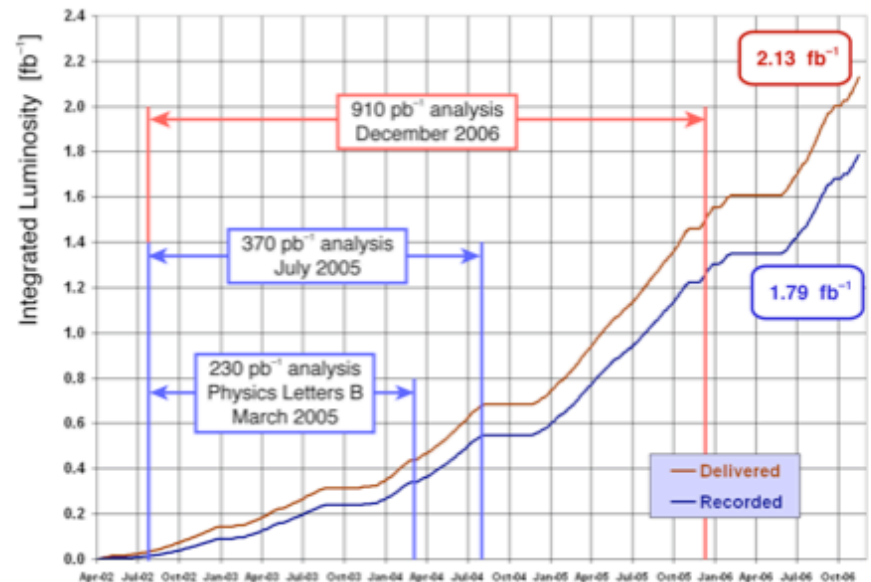


# Tevatron luminosity



## Run II Integrated Luminosity

Apr 2002 – Dec 2006



Many thanks to the Accelerator Division

# Comparison

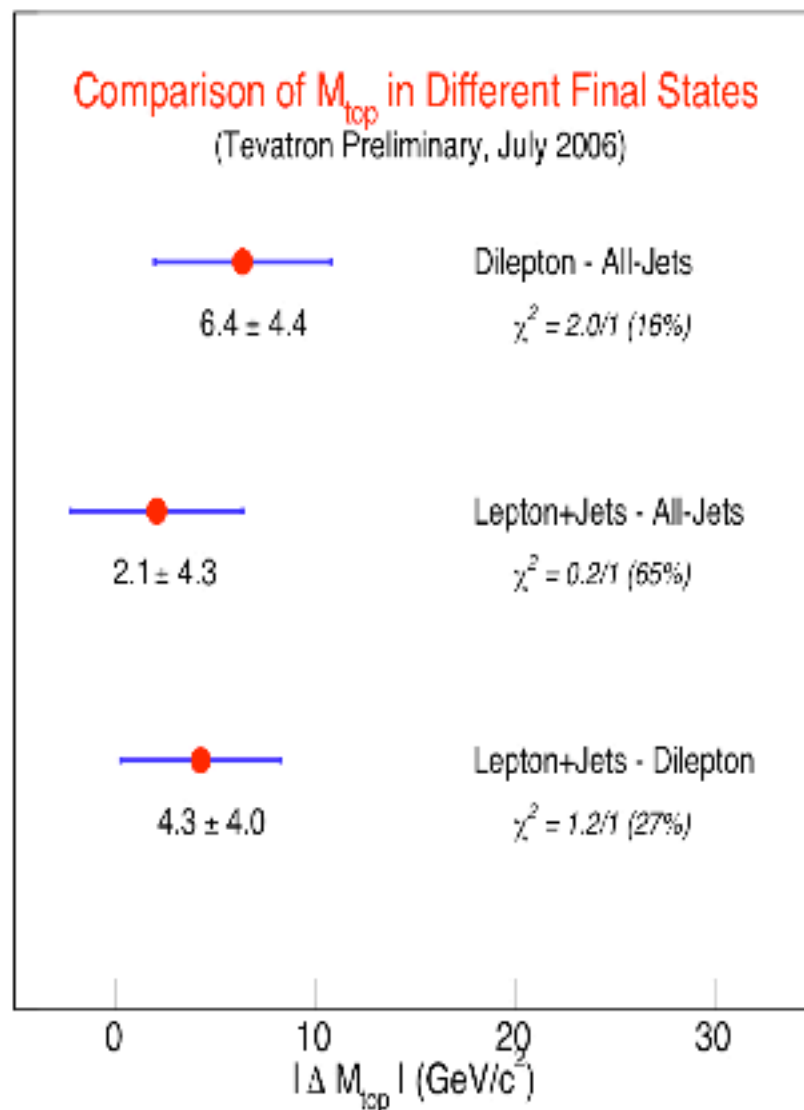
- Are the channels consistent?

$$\begin{aligned} M_{\text{top}}(\text{All Jets}) (*) &= 173.4 \pm 4.3 \text{ GeV}/c^2 \\ M_{\text{top}}(\text{Dilepton}) &= 167.0 \pm 4.3 \text{ GeV}/c^2 \\ M_{\text{top}}(\text{Lepton+Jets}) &= 171.3 \pm 2.2 \text{ GeV}/c^2 \end{aligned}$$

- We compare them taking into account their correlated systematic uncertainties

⇒ Determination of  $M_{\text{top}}$  from the 3 different channels is consistent with one another

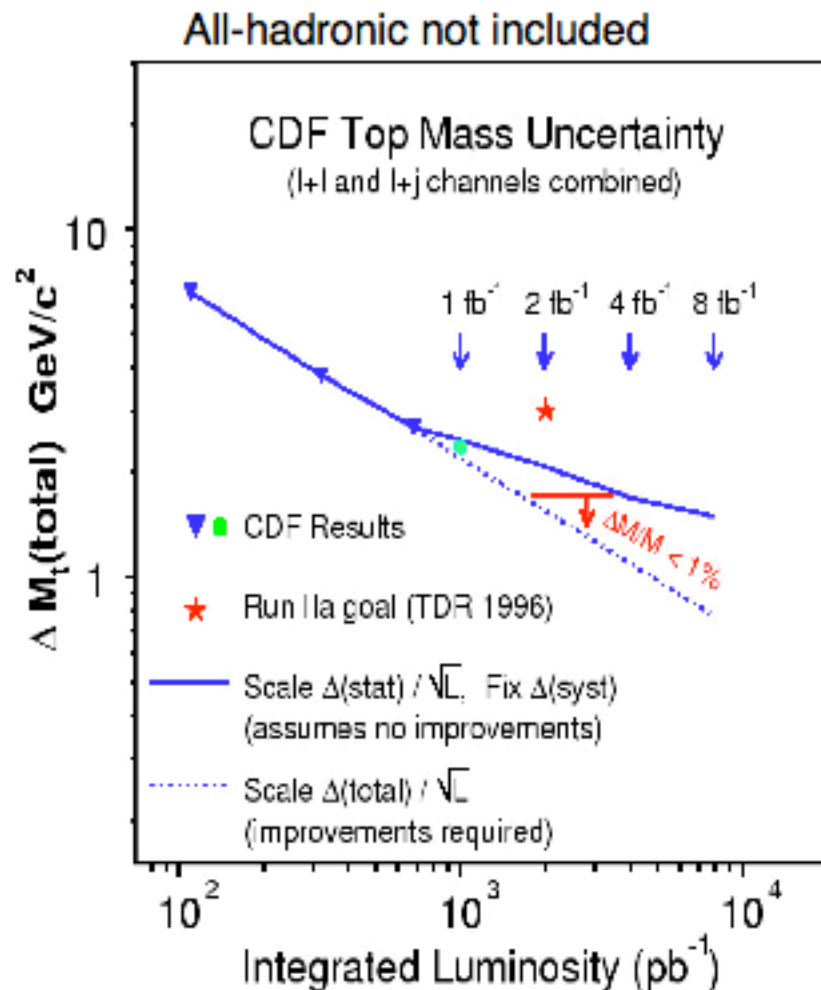
(\*) not including latest CDF 2D all-hadronic result





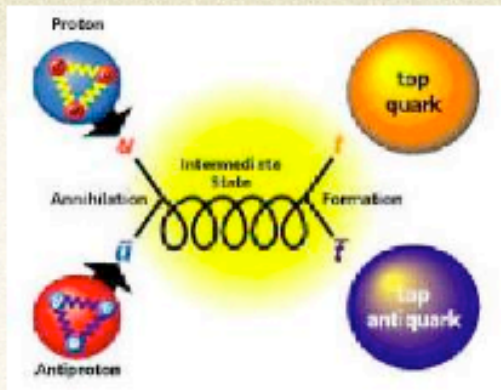
# Future of Top Mass Measurements

- New results are better than our predictions 6 months ago
- Add JES to all-hadronic channel makes sensitivity comparable to lepton+jets
- $D\bar{D}$  has similar sensitivity (new results with  $1 \text{ fb}^{-1}$  coming soon)
- We expect to achieve an uncertainty of  $<1\%$  in the next years



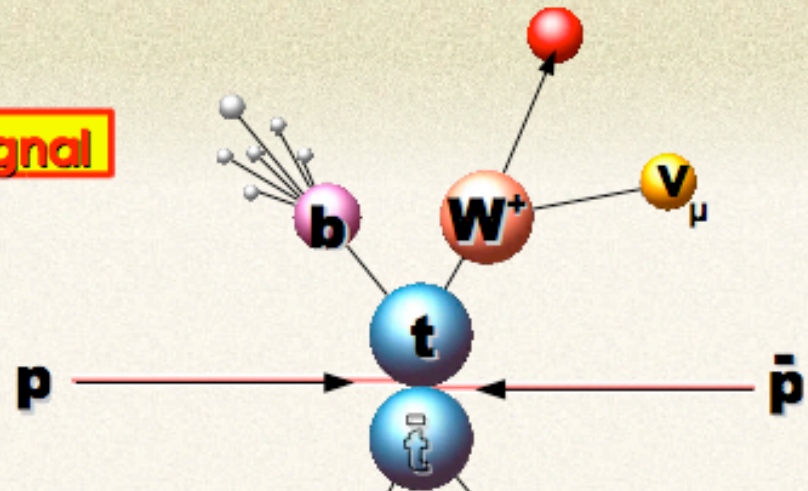
# Characterization of Dilepton Events

## strong production of top pairs

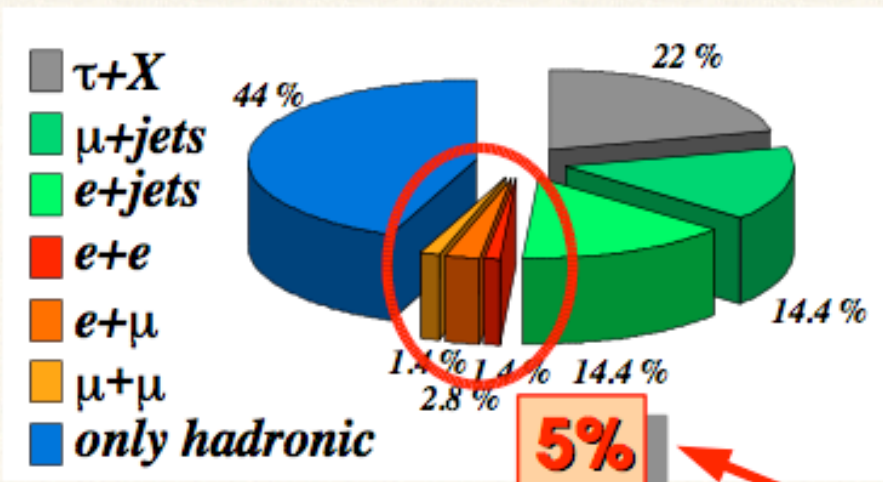


85%

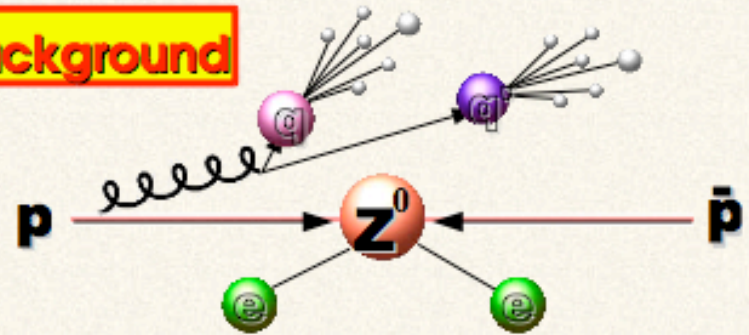
signal



## final states



background



small background  $\Rightarrow$  precise measurement in future

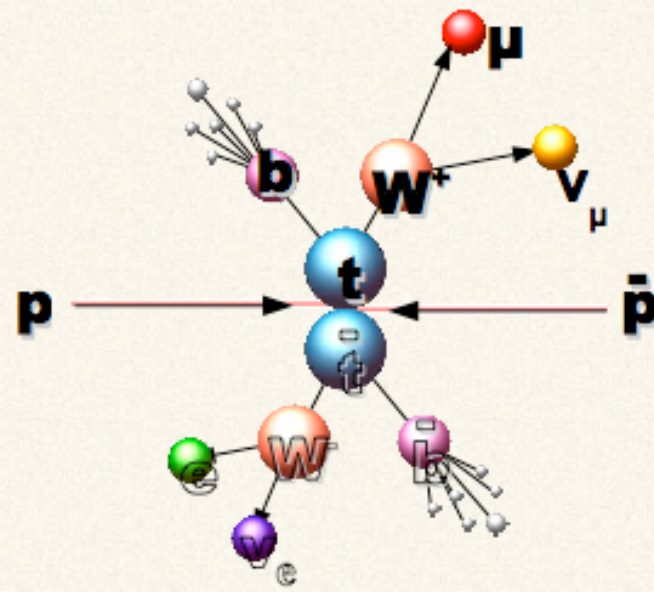
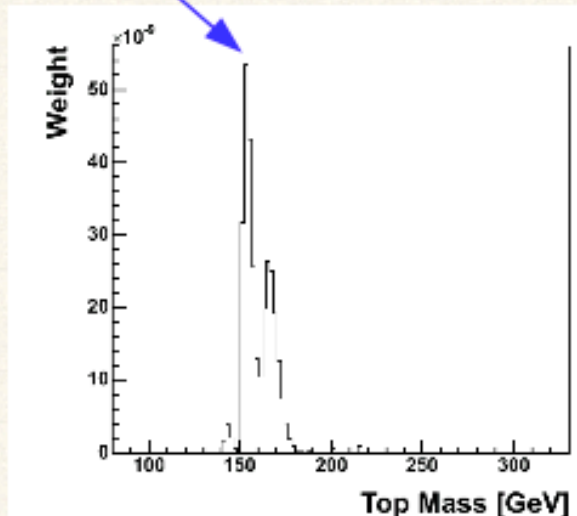
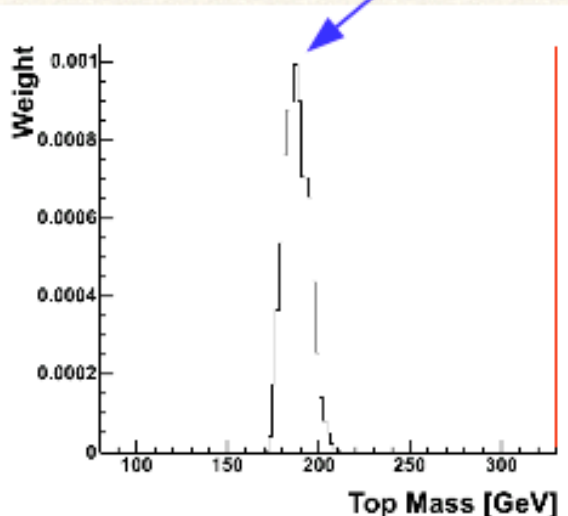
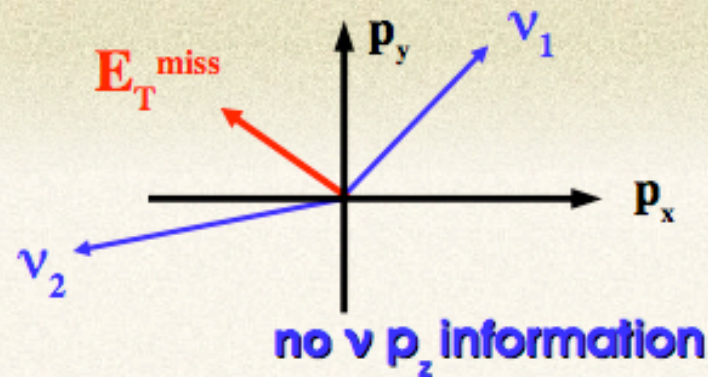
# Characterization of Dilepton Events

## Problem:

- kinematics is underconstrained due to 2 neutrinos
- multiple solutions: which jet or  $(l, \nu)$  pair belongs to which top or anti-top quark?

→ apply weight  $W(m_t)$  to each event:  
**neutrino/matrix weighting algorithm**

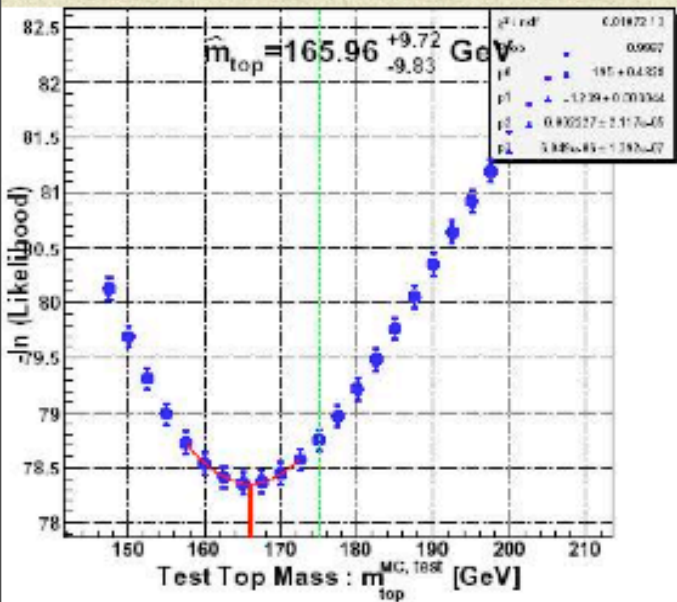
most likely value for top mass



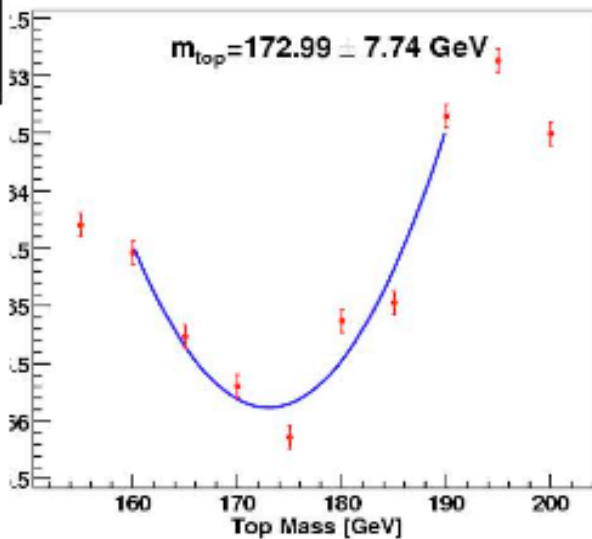
**2 example events**



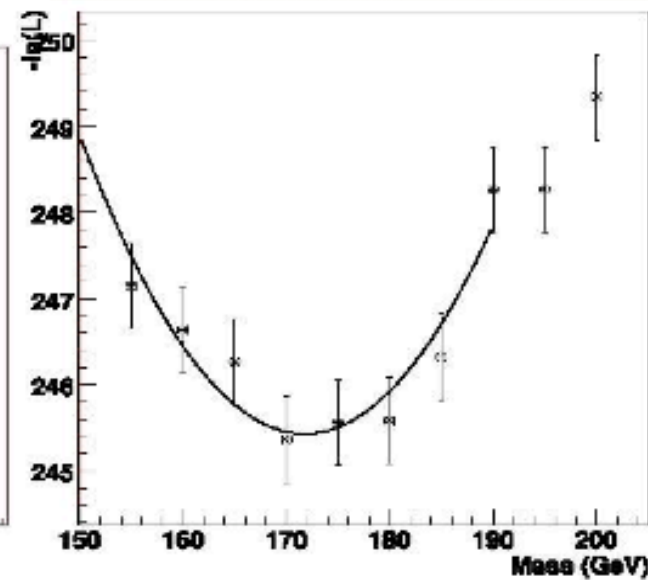
## Maximum Method



## Binned Method



## Moments Method



**Maximum Method:**  $m_{top} = 165.7 \pm 9.7$  (stat.)  $^{+4.4}_{-4.7}$  (syst.) GeV

**Binned Method:**  $m_{top} = 173.6 \pm 6.7$  (stat.)  $^{+5.1}_{-4.0}$  (syst.) GeV

**Moments Method:**  $m_{top} = 171.6 \pm 7.9$  (stat.)  $^{+5.1}_{-4.0}$  (syst.) GeV

(835 pb<sup>-1</sup>,  $e\mu$  channel only)



# Dilepton Matrix Element



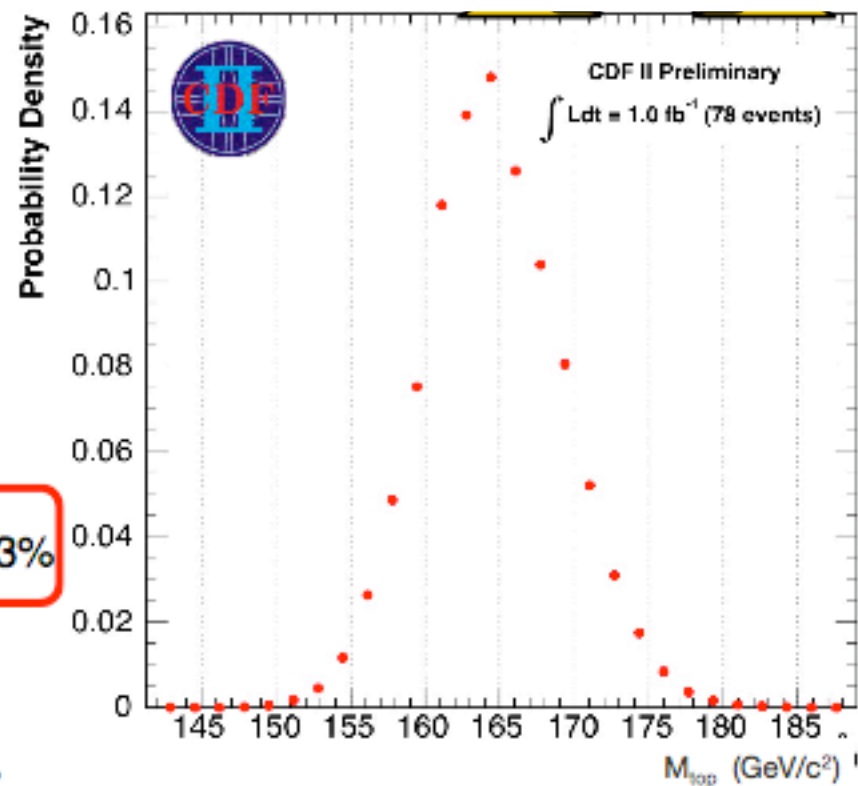
- Probability density calculated for  $t\bar{t}$  and 3 of the major backgrounds

- Using 1030 pb<sup>-1</sup> and 78 candidates CDF measures

$$M_{top} = 164.5 \pm 3.9(\text{stat.}) \pm 3.5(\text{JES}) \pm 1.7(\text{syst.}) \text{ GeV}/c^2 \quad 3.3\%$$

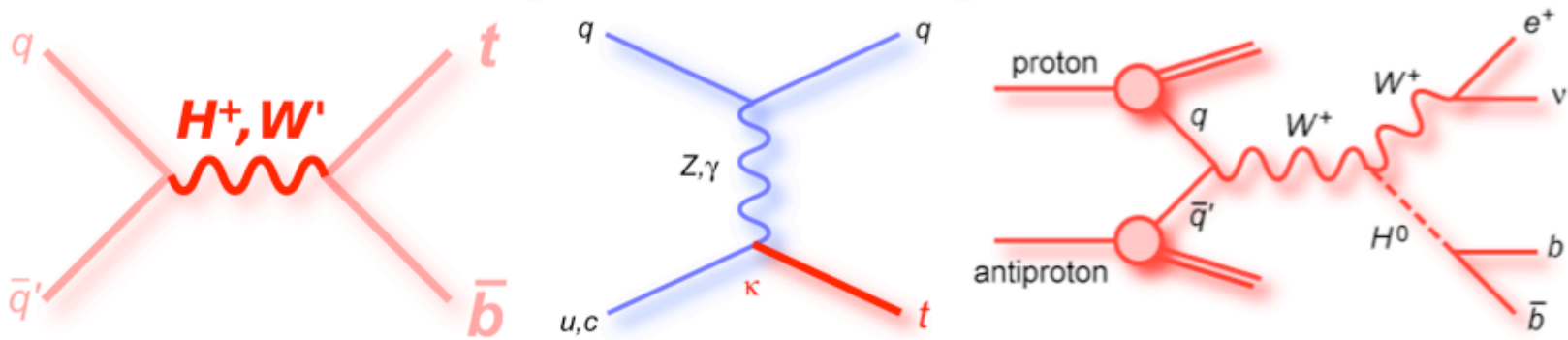
- Cross-check result requiring b-tagging

$$M_{top} = 167.3 \pm 4.6(\text{stat.}) \pm 3.3(\text{JES}) \pm 1.9(\text{syst.}) \text{ GeV}/c^2$$



# Motivation

- Directly measure  $|V_{tb}|$  (more later)
- Cross sections sensitive to new physics:
  - s-channel: resonances (heavy  $W'$  boson, charged Higgs boson  $H^\pm$ , Kaluza-Klein excited  $W_{KK}$ , etc...)
  - t-channel: flavour-changing neutral currents ( $t - Z/\gamma/g - c$  couplings)
  - Fourth generation of quarks
- Source of polarized top quarks. Spin correlations measurable in decay products
- Important background to  $WH$  associated Higgs production
  - if the tools don't work for single top, forget about the Higgs
- Test of techniques to extract a small signal out of a large background



# Event selection – S/B

Percentage of single top *tb+tb* selected events and S:B ratio (white squares = no plans to analyze)

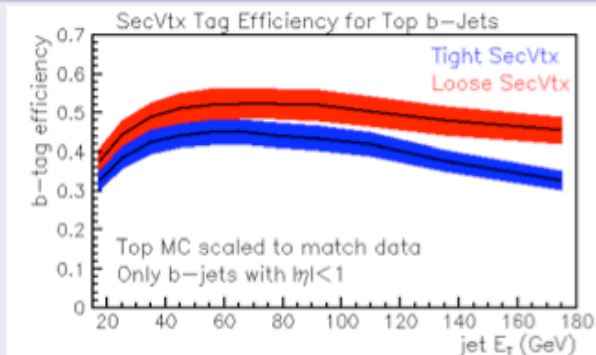
Electron + Muon	1 jet	2 jets	3 jets	4 jets	≥ 5 jets
0 tags	10% 1 : 3,200	25% 1 : 390	12% 1 : 300	3% 1 : 270	1% 1 : 230
1 tag	6% 1 : 100	21% 1 : 20	11% 1 : 25	3% 1 : 40	1% 1 : 53
2 tags		3% 1 : 11	2% 1 : 15	1% 1 : 38	0% 1 : 43



# CDF b tagging

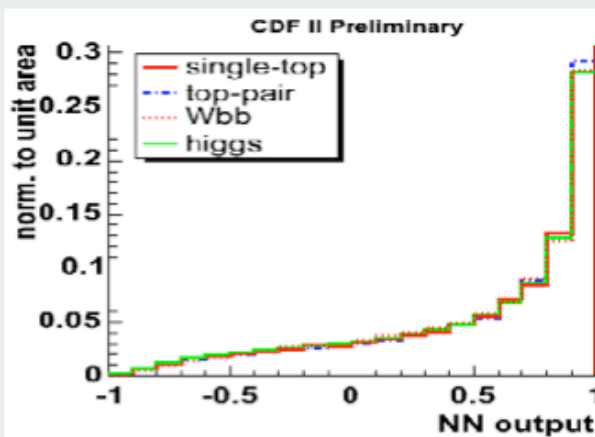
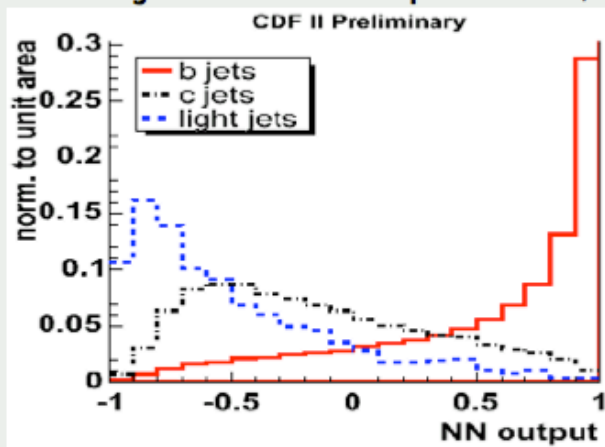
## Secondary vertex tagging

- Long lifetime of  $B$  hadrons
- Travel several mm before decaying
- Signature: displaced secondary vertex tagger
- Tagging efficiency per jet  $\sim 40\%$



## Jet flavour separation

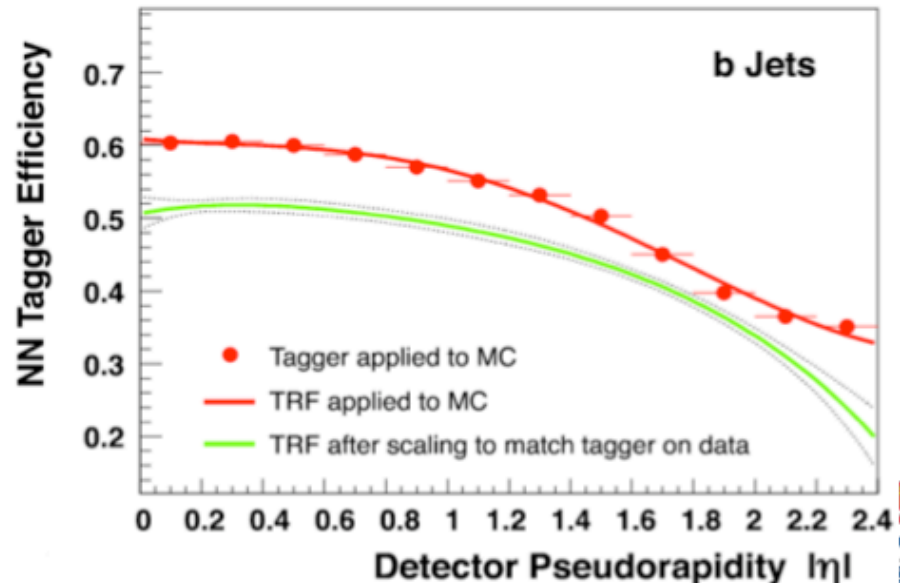
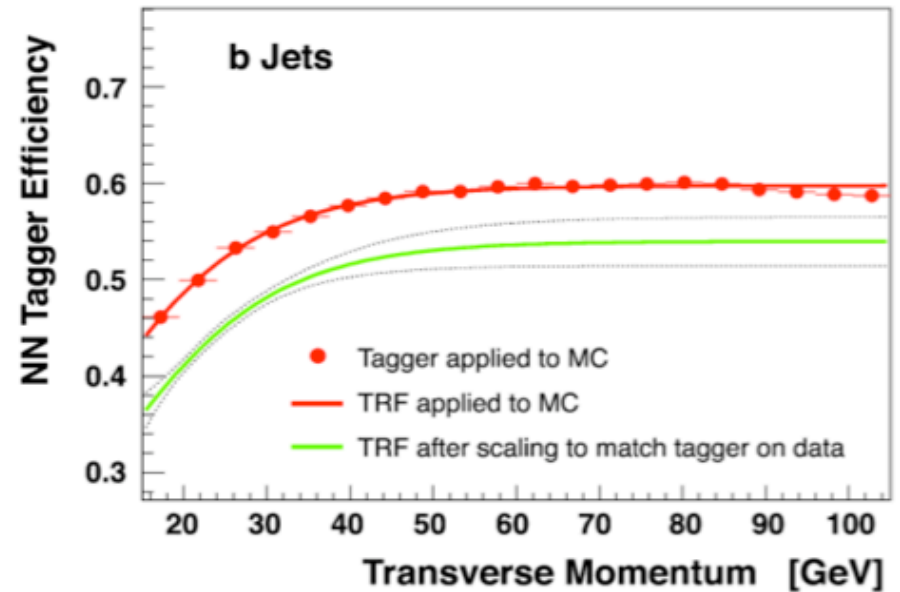
- Second stage: improve separation with 25-input neural network
- Applied on jets  $b$ -tagged with secondary vertex
- Good jet flavour separation, independent of  $b$ -jet source





# DØ b tagging

- NN trained on 7 input variables from existing taggers.
- Much improved performance!
  - fake rate reduced by 1/3 for same  $b$  efficiency relative to previous tagger
  - smaller systematic uncertainties
- Tag Rate Functions (TRFs) in  $\eta$ ,  $p_T$ ,  $z$ -PV applied to MC
- Operating point:
  - $b$ -jet efficiency  $\sim 50\%$
  - $c$ -jet efficiency  $\sim 10\%$
  - light jet efficiency  $\sim 0.5\%$



# Systematic uncertainties – CDF

CDF RunII Preliminary,  $L=955\text{pb}^{-1}$

Single Top	Rate Variations	Shape Variations
Jet Energy Scale	✓	✓
Initial State Radiation	✓	✓
Final State Radiation	✓	✓
Parton Dist. Function	✓	✓
Monte Carlo Generator	✓	
Efficiencies / b-tagging SF	✓	
Luminosity	✓	
<b>Total Rate Uncertainty</b>	<b>10.5%</b>	<b>N/A</b>

Backgrounds	Rate Variations	Shape Variations
Jet Energy Scale	✓	✓
Neural Net b-tagger		✓
Mistag Model		✓
Non-W Model		✓
$Q^2$ Scale in Alpgen MC		✓

Background	Rate Uncertainty
W+bottom	28%
W+charm	28%
Mistag	15%
ttbar	23%

- Rate and shape uncertainties included as nuisance parameters in analyses

# Systematic uncertainties – DØ

- Assigned per background, jet multiplicity, lepton flavour and number of tags
- Uncertainties that affect both normalisation and shapes: jet energy scale and tag rate functions (*b*-tagging parameterisation)
- All uncertainties sampled during limit-setting phase

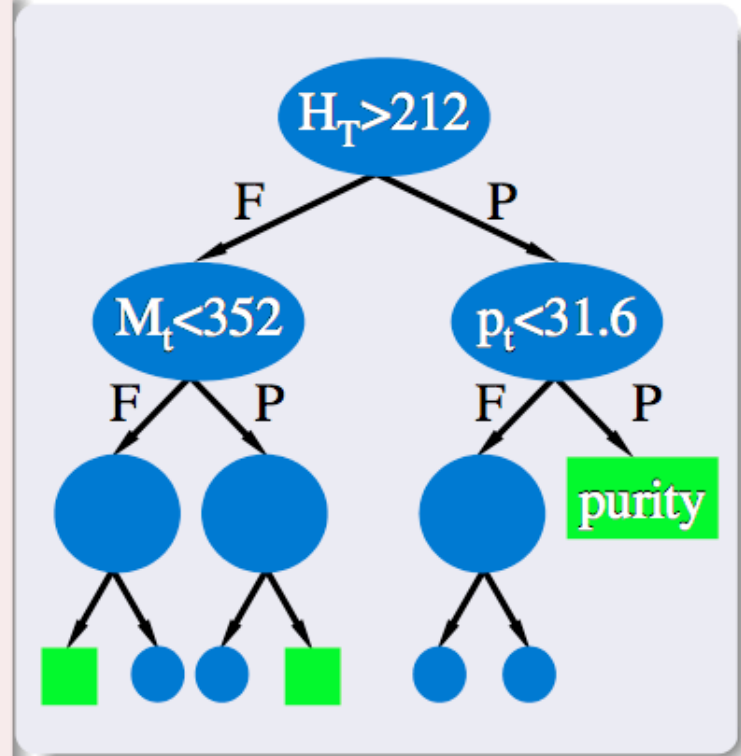
## Relative systematic uncertainties

$t\bar{t}$ cross section	18%	Primary vertex	3%
Luminosity	6%	<i>e</i> reco * ID	2%
Electron trigger	3%	<i>e</i> trackmatch & likelihood	5%
Muon trigger	6%	$\mu$ reco * ID	7%
Jet energy scale	wide range	$\mu$ trackmatch & isolation	2%
Jet efficiency	2%	$\epsilon_{\text{real}-e}$	2%
Jet fragmentation	5–7%	$\epsilon_{\text{real}-\mu}$	2%
Heavy flavor ratio	30%	$\epsilon_{\text{fake}-e}$	3–40%
Tag-rate functions	2–16%	$\epsilon_{\text{fake}-\mu}$	2–15%

# Decision trees

- Machine-learning technique, widely used in social sciences
- Idea: recover events that fail criteria in cut-based analysis

- Start with all events = first node
  - sort all events by each variable
  - for each variable, find splitting value with best separation between two children (mostly signal in one, mostly background in the other)
  - select variable and splitting value with best separation, produce two branches with corresponding events ((F)ailed and (P)assed cut)
- Repeat recursively on each node
- Splitting stops: terminal node = leaf



- Run testing events and data through tree to derive limits
- DT output = leaf purity, close to 1 (0) for signal (bkg)

Ref: Breiman *et al*, "Classification and Regression Trees", Wadsworth (1984)

# Boosting a decision tree

## Boosting

- Recent technique to improve performance of a weak classifier
- Recently used on decision trees by GLAST and MiniBooNE
- Basic principal on DT:
  - train a tree  $T_k$
  - $T_{k+1} = \text{modify}(T_k)$

## AdaBoost algorithm

- Adaptive boosting
- Check which events are misclassified by  $T_k$
- Derive tree weight  $\alpha_k$
- Increase weight of misclassified events
- Train again to build  $T_{k+1}$
- Boosted result of event  $i$ :  
$$T(i) = \sum_{n=1}^{N_{\text{tree}}} \alpha_n T_n(i)$$

- Averaging  $\Rightarrow$  dilutes piecewise nature of DT
- Usually improves performance

Ref: Freund and Schapire, "Experiments with a new boosting algorithm", in *Machine Learning: Proceedings of the Thirteenth International Conference*, pp 148-156 (1996)

# Decision trees at DØ

## DT choices

- 1/3 of MC for training
- AdaBoost parameter  $\beta = 0.2$
- 20 boosting cycles
- Signal leaf if purity  $> 0.5$
- Minimum leaf size = 100 events
- Same total weight to signal and background to start
- Goodness of split - Gini factor

## Input variables

- Used 49 variables (object and event kinematics, angular correlations)
- Adding variables does not degrade performance
- Tested shorter lists: lost some sensitivity
- Same list used for all channels

## Analysis strategy

- Train 36 separate trees:  $(s, t, s + t) \times (e, \mu) \times (2, 3, 4 \text{ jets}) \times (1, 2 \text{ tags})$
- For each signal train against the sum of backgrounds

# Matrix element method

- Pioneered by DØ top mass analysis. Now used in search
- Use the 4-vectors of all reconstructed leptons and jets
- Use matrix elements of main signal and background diagrams to compute an event probability density for signal and background hypotheses
- Encoded in properly normalized differential cross section for process  $S$ :

$$P_S(\vec{x}) = \frac{1}{\sigma_S} d\sigma_S(\vec{x}), \quad \sigma_S = \int d\sigma_S(\vec{x})$$

- Only a limited number of Feynman diagrams are used. Sensitivity would increase (but so does computation time) if more diagrams were included. In particular, no  $t\bar{t}$  diagrams are computed (serious limitation for  $>2$  jets)

# Matrix element discriminants

## DØ discriminants

$$D_s(\vec{x}) = P(S|\vec{x}) = \frac{P_{signal}(\vec{x})}{P_{signal}(\vec{x}) + P_{bkg}(\vec{x})}$$

$$P_{bkg}^{2jets}(\vec{x}) = c_{Wbb}P_{Wbb}(\vec{x}) + c_{Wcg}P_{Wcg}(\vec{x}) + c_{Wgg}P_{Wgg}(\vec{x})$$

$$P_{bkg}^{3jets}(\vec{x}) = P_{Wbbg}(\vec{x})$$

- $c_{Wbb}$ ,  $c_{Wcg}$  and  $c_{Wgg}$  are in principle the relative fractions of each background
- optimized for each channel to increase sensitivity

## CDF discriminant

$$EPD = \frac{b \cdot P_{signal}}{b \cdot P_{signal} + b \cdot P_{Wbb} + (1 - b)P_{Wcc} + (1 - b)P_{Wcj}}$$

- $b$  is the neural network  $b$ -tagger output converted to probability



# Likelihood method (CDF)

- Likelihood for a vector of measurements  $\vec{x} = x_i$ :

$$\mathcal{L}(\vec{x}) = \frac{\mathcal{P}_{signal}(\vec{x})}{\mathcal{P}_{signal}(\vec{x}) + \sum \mathcal{P}_{background}(\vec{x})}, \quad \mathcal{P}(\vec{x}) = \prod_i^{N_{variables}} P(x_i)$$

$P(x_i)$  = normalized  $x_i$  variable distribution

- Four backgrounds:  $Wbb$ ,  $t\bar{t}$ ,  $Wcc/Wc$ , mistags

## t-channel LF Variables:

- total transverse energy:  $H_T$
- $M_{lvb}$  (neutrino  $p_z$  from kin. fitter)
- $\text{Cos}\theta(\text{lepton, light jet})$  in top decay frame
- $Q_{\text{lepton}} * \eta_{\text{untagged jet}}$  aka  $Q_x\text{Eta}$
- $m_{j1j2}$
- $\log(\text{ME}_{t\text{chan}})$  from MADGRAPH
- Neural Network b-tagger
- LF=0.01 for double tagged events

## s-channel LF Variables:

- $M_{lvb}$
- $\log(H_T * M_{lvb})$
- $E_T(\text{jet1})$
- $\log(\text{ME}_{t\text{chan}})$
- $H_T$
- Neural Network b-tagger

# Neural network – Bayesian neural networks

## Neural network (CDF)

- Three-layer perceptrons using NeuroBayes
- Continuous output between -1 (bkg-like) and +1 (signal-like)
- 26 input variables
- Three networks: tb, tqb and tb+tqb and signal

## Bayesian neural networks (DØ)

- Instead of choosing one set of weights, find posterior probability density over all possible weights
- Averaging over many networks weighted by the probability of each network given the training data
- Less prone to overtraining
- For details see:  
<http://www.cs.toronto.edu/~radford/fbm.software.html>
- Use 24 variables (subset of DT variables)

# DØ analysis validation

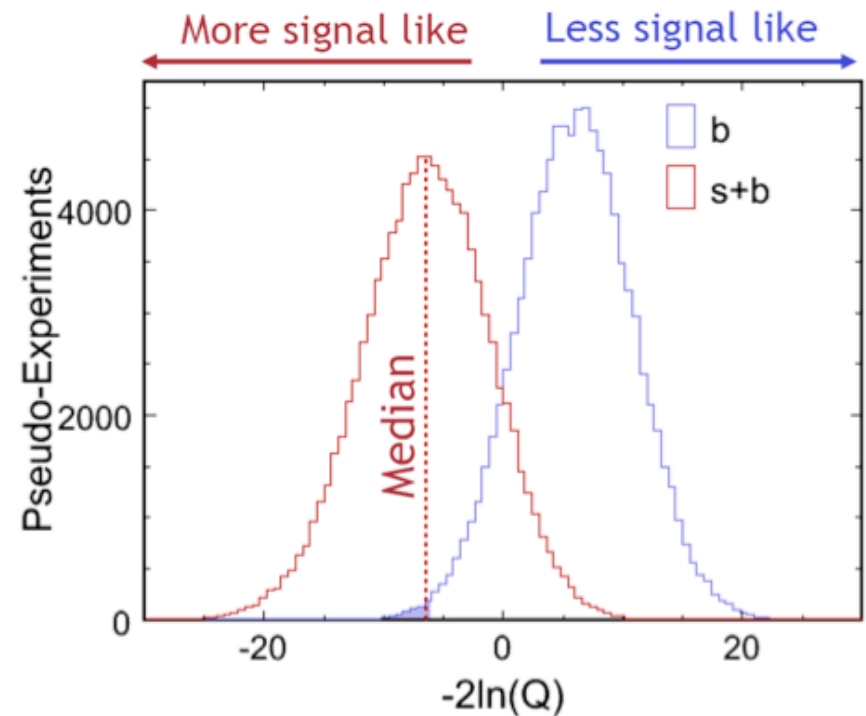
- To verify that all of this machinery is working properly we test with many sets of pseudo-data.
- Wonderful tool to test analysis methods! Run DØ experiment 1000s of times!
- Generated ensembles:
  - 0-signal ensemble ( $s + t \sigma = 0$  pb)
  - SM ensemble ( $s + t \sigma = 2.9$  pb)
  - “Mystery” ensembles to test analyzers ( $s + t \sigma = ??$  pb)
  - Ensembles at measured cross section ( $s + t \sigma = \text{measured}$ )
  - A high luminosity ensemble
- All analyses achieved linear response to varying input cross sections

# Sensitivity determination at CDF

- Using the CLs method developed at LEP
- Compare two models at a time
- Test statistic:

$$Q = \frac{L(\text{data}|s + b)}{L(\text{data}|b)}$$

- Systematic uncertainties included in pseudo-experiments
- **Expected sensitivity**: median p-value



Likelihood	median p-value = 2.3%	(2.0 $\sigma$ )
Matrix element	median p-value = 0.6%	(2.5 $\sigma$ )
Neural network	median p-value = 0.5%	(2.6 $\sigma$ )

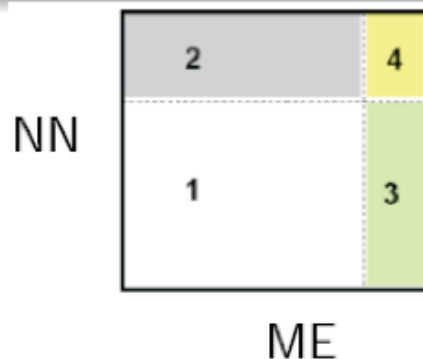
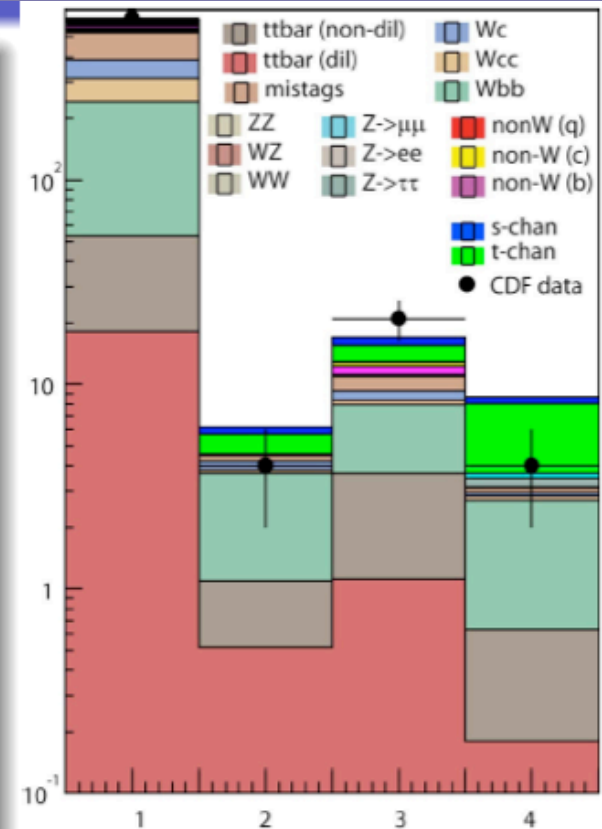
# Sensitivity determination at DØ

- Use the 0-signal ensemble:
  - use pool of weighted signal+bkg events
  - fluctuate relative and total yields in proportion to syst. errors
  - randomly sample from a Poisson distribution about total yield
  - generate a set of pseudo data
  - pass the pseudo-data through the full analysis
- **Expected p-value:** fraction of 0-signal pseudo-datasets in which we measure at least 2.9 pb (SM single top cross section)
- **Observed p-value:** fraction of 0-signal pseudo-datasets in which we measure at least the observed cross section.

Boosted decision trees	p-value = 1.9%	(2.1 $\sigma$ )
Matrix element	p-value = 3.7%	(1.8 $\sigma$ )
Bayesian neural networks	p-value = 9.7%	(1.3 $\sigma$ )

# CDF observed results – Compatibility

- CDF spent great deal of time (6 months) and effort understanding if the different results are something more than a statistical fluctuation.
- Eliminated possibility of obvious and even subtle bugs
- 6-discriminant compatibility coming soon
- Now investigating if features of the MC modeling affect one analysis more than the other.
- Analysing more data should shed some light



- Bin 1:  $NN < 0.8 \ \&\& \ EPD < 0.9$
- Bin 2:  $NN > 0.8 \ \&\& \ EPD < 0.9$
- Bin 3:  $NN < 0.8 \ \&\& \ EPD > 0.9$
- Bin 4:  $NN > 0.8 \ \&\& \ EPD > 0.9$

# DØ BNN and ME s+t observed results

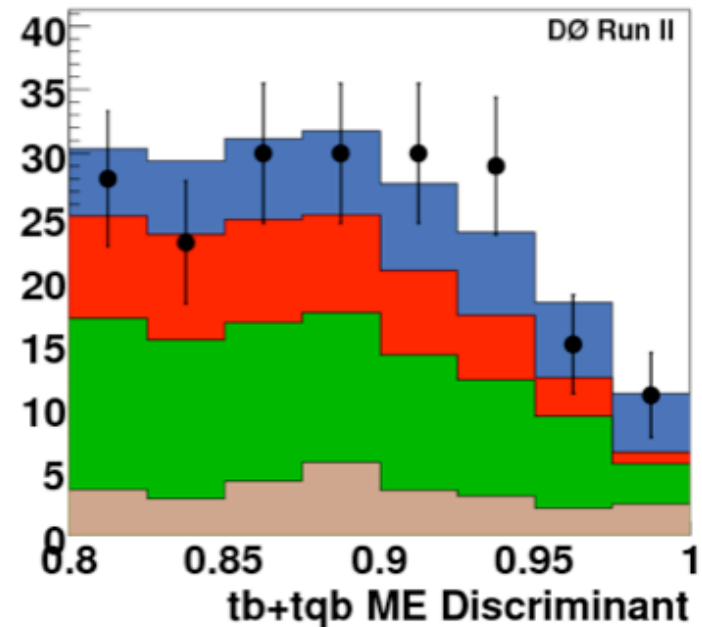
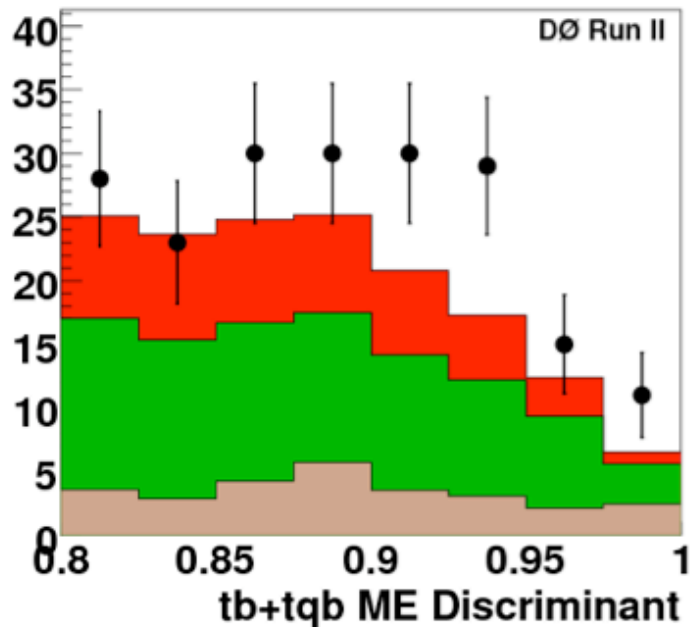
## Bayesian NN

$\sigma = 5.0 \pm 1.9 \text{ pb}$   
p-value = 0.89% ( $2.4\sigma$ )

## Matrix element

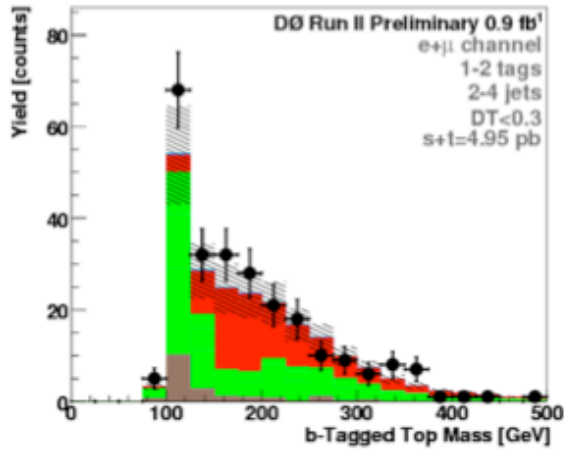
$\sigma = 4.6^{+1.8}_{-1.5} \text{ pb}$   
p-value = 0.21% ( $2.9\sigma$ )

- ME discriminant output, with and without signal content (all channels combined)

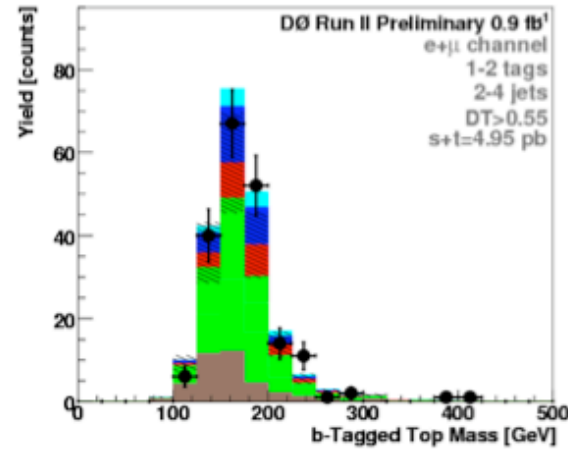


# DØ boosted decision tree event characteristics

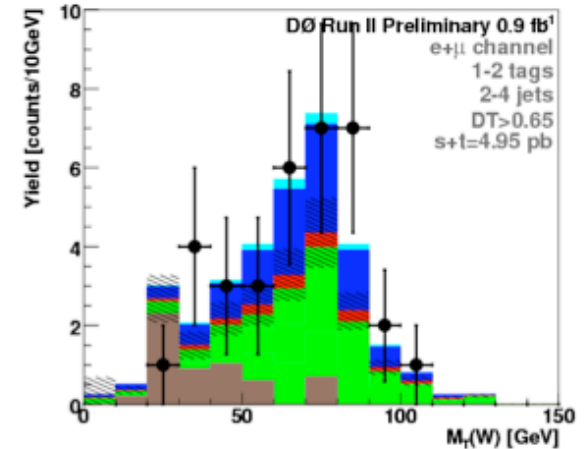
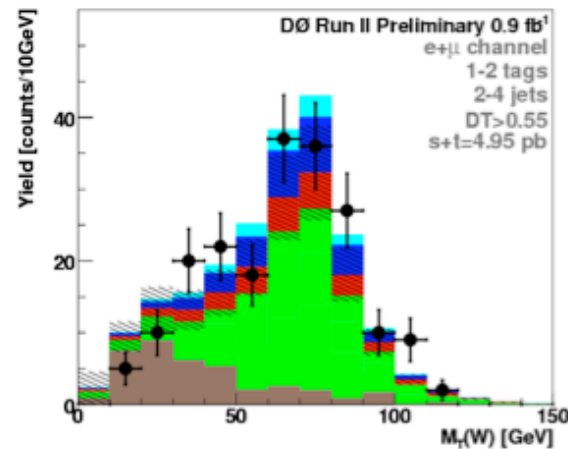
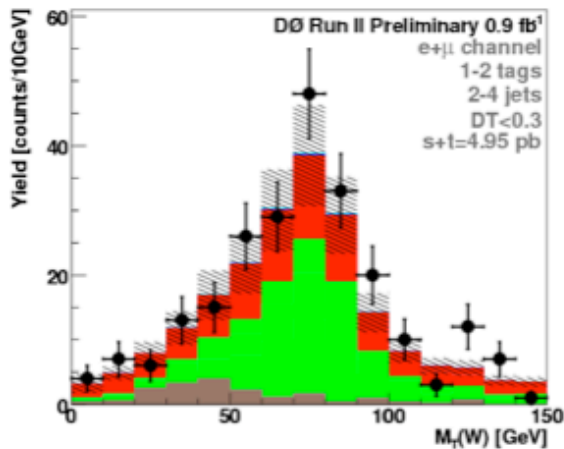
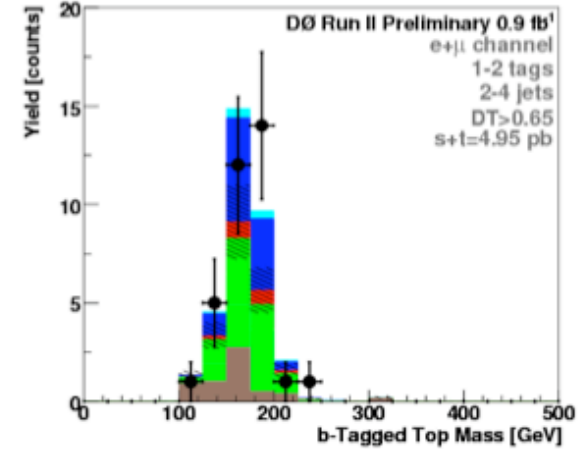
$DT < 0.3$



$DT > 0.55$



$DT > 0.65$





# Splitting a node

## Impurity $i(t)$

- maximum for equal mix of signal and background
- symmetric in  $p_{signal}$  and  $p_{background}$
- minimal for node with either signal only or background only
- strictly concave  $\Rightarrow$  reward purer nodes

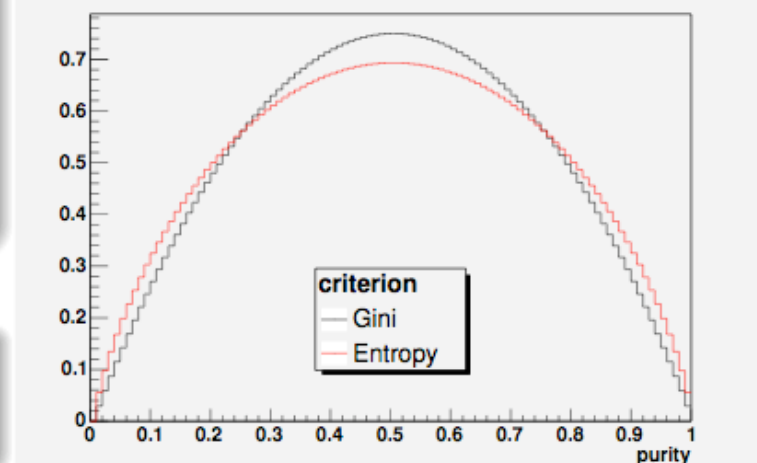
- Decrease of impurity for split  $s$  of node  $t$  into children  $t_L$  and  $t_R$  (goodness of split):  
$$\Delta i(s, t) = i(t) - p_L \cdot i(t_L) - p_R \cdot i(t_R)$$
- Aim: find split  $s^*$  such that:

$$\Delta i(s^*, t) = \max_{s \in \{\text{splits}\}} \Delta i(s, t)$$

- Maximizing  $\Delta i(s, t) \equiv$  minimizing overall tree impurity

## Examples

$$Gini = 1 - \sum_{i=s,b} p_i^2 = \frac{2sb}{(s+b)^2}$$
$$entropy = - \sum_{i=s,b} p_i \log p_i$$



# Decision trees – 49 variables

## Object Kinematics

$p_T(\text{jet1})$   
 $p_T(\text{jet2})$   
 $p_T(\text{jet3})$   
 $p_T(\text{jet4})$   
 $p_T(\text{best1})$   
 $p_T(\text{notbest1})$   
 $p_T(\text{notbest2})$   
 $p_T(\text{tag1})$   
 $p_T(\text{untag1})$   
 $p_T(\text{untag2})$

## Angular Correlations

$\Delta R(\text{jet1}, \text{jet2})$   
 $\cos(\text{best1}, \text{lepton})_{\text{besttop}}$   
 $\cos(\text{best1}, \text{notbest1})_{\text{besttop}}$   
 $\cos(\text{tag1}, \text{alljets})_{\text{alljets}}$   
 $\cos(\text{tag1}, \text{lepton})_{\text{btaggedtop}}$   
 $\cos(\text{jet1}, \text{alljets})_{\text{alljets}}$   
 $\cos(\text{jet1}, \text{lepton})_{\text{btaggedtop}}$   
 $\cos(\text{jet2}, \text{alljets})_{\text{alljets}}$   
 $\cos(\text{jet2}, \text{lepton})_{\text{btaggedtop}}$   
 $\cos(\text{lepton}, Q(\text{lepton}) \times z)_{\text{besttop}}$   
 $\cos(\text{lepton}, \text{besttopframe})_{\text{besttopCMframe}}$   
 $\cos(\text{lepton}, \text{btaggedtopframe})_{\text{btaggedtopCMframe}}$   
 $\cos(\text{notbest}, \text{alljets})_{\text{alljets}}$   
 $\cos(\text{notbest}, \text{lepton})_{\text{besttop}}$   
 $\cos(\text{untag1}, \text{alljets})_{\text{alljets}}$   
 $\cos(\text{untag1}, \text{lepton})_{\text{btaggedtop}}$

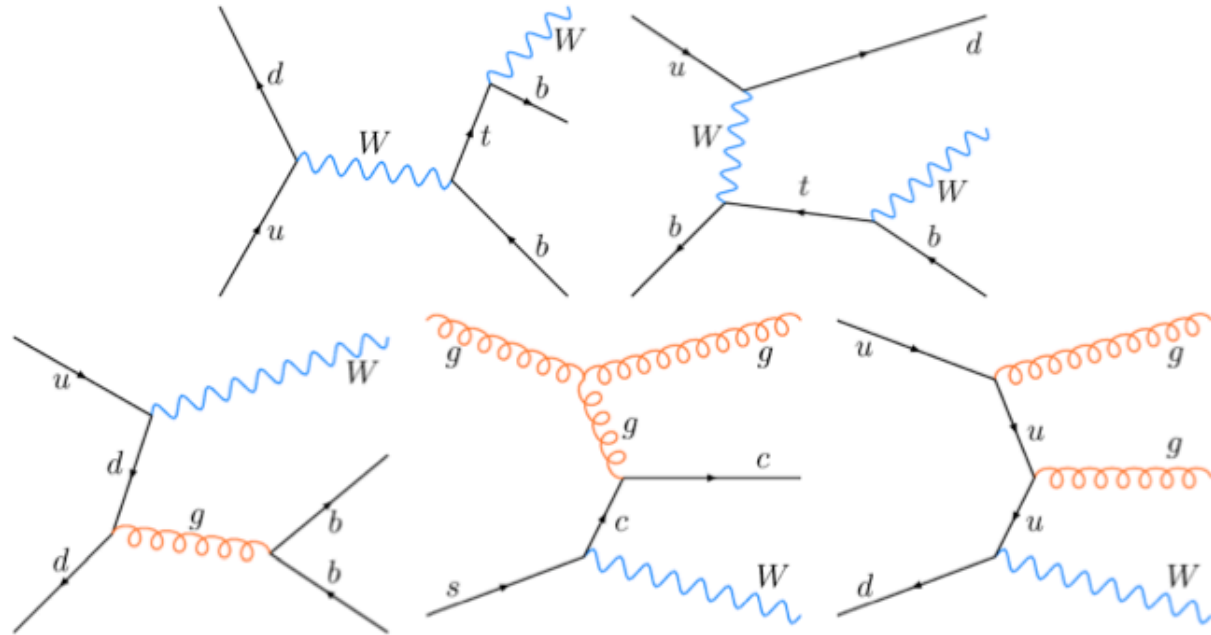
## Event Kinematics

Aplanarity(alljets,  $W$ )  
 $M(W, \text{best1})$  ("best" top mass)  
 $M(W, \text{tag1})$  ("b-tagged" top mass)  
 $H_T(\text{alljets})$   
 $H_T(\text{alljets} - \text{best1})$   
 $H_T(\text{alljets} - \text{tag1})$   
 $H_T(\text{alljets}, W)$   
 $H_T(\text{jet1}, \text{jet2})$   
 $H_T(\text{jet1}, \text{jet2}, W)$   
 $M(\text{alljets})$   
 $M(\text{alljets} - \text{best1})$   
 $M(\text{alljets} - \text{tag1})$   
 $M(\text{jet1}, \text{jet2})$   
 $M(\text{jet1}, \text{jet2}, W)$   
 $M_T(\text{jet1}, \text{jet2})$   
 $M_T(W)$   
Missing  $E_T$   
 $p_T(\text{alljets} - \text{best1})$   
 $p_T(\text{alljets} - \text{tag1})$   
 $p_T(\text{jet1}, \text{jet2})$   
 $Q(\text{lepton}) \times \eta(\text{untag1})$   
 $\sqrt{\hat{s}}$   
Sphericity(alljets,  $W$ )

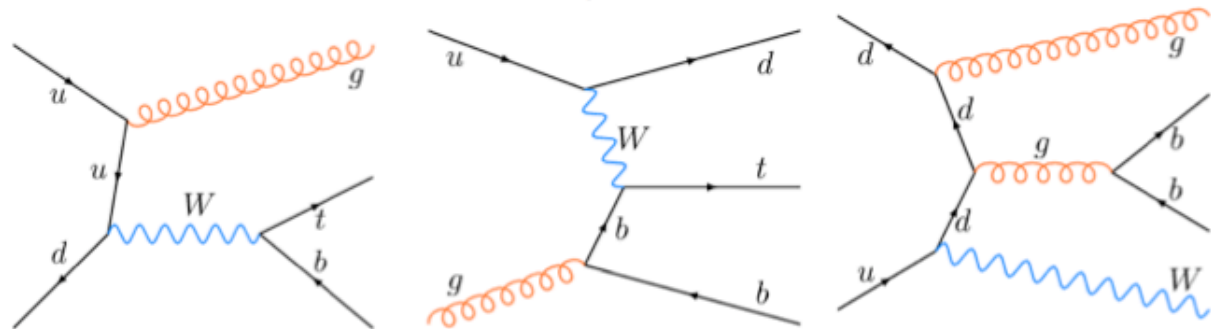
- Adding variables does not degrade performance
- Tested shorter lists, lose some sensitivity
- Same list used for all channels

# Matrix element method – DØ diagrams

2-jets:



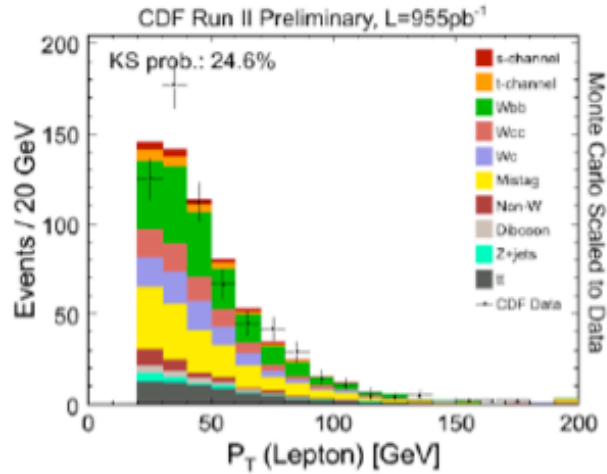
3-jets:



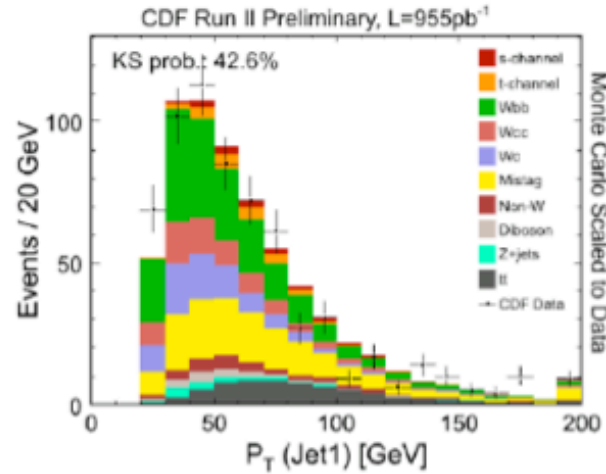
# CDF ME inputs (B. Stelzer, R. Walny)

- Input to the Matrix Element Analysis are the measured four-vectors of the Lepton, Jet1 and Jet2 in the W+2jets data ( $\geq 1$  b-tagged jet)

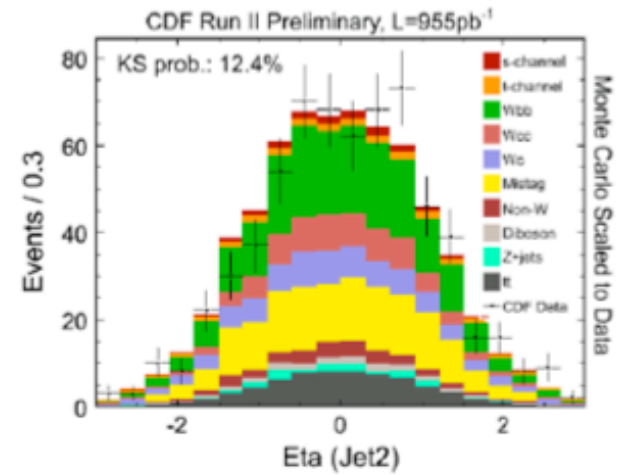
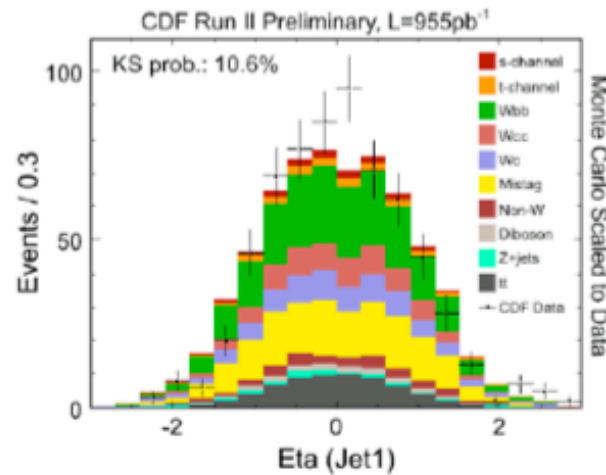
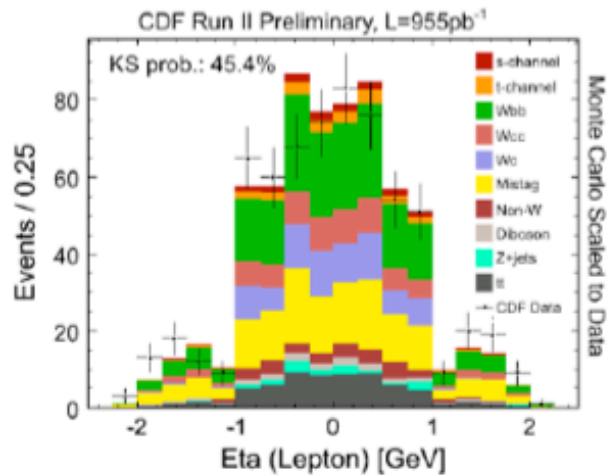
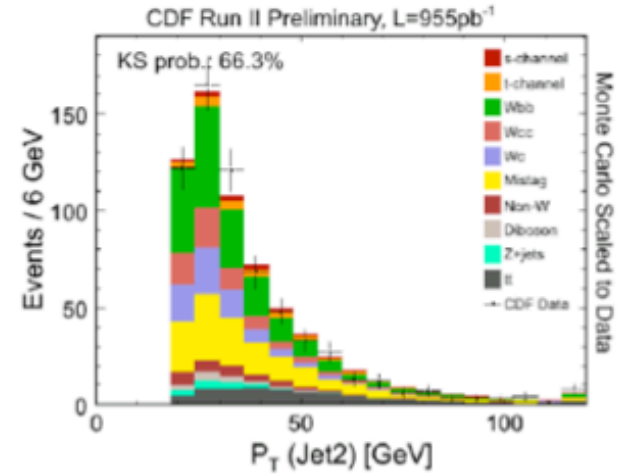
Lepton



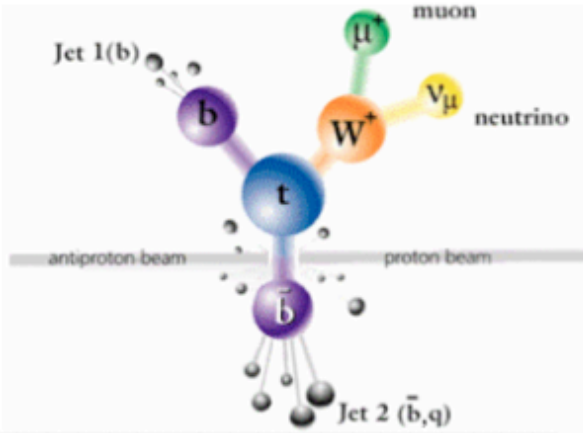
Jet1



Jet2



# Matrix element method – Probability



Single Top kinematic quantities:

- $2(\text{initial}) + 12(\text{final}) = 14$  **degrees of freedom**
- Assume leptons and angles well measured
- $3(l) + 4(\text{angle}) + 3(P_{\text{in}} = P_{\text{fin}}) + 1(E_{\text{in}} = E_{\text{fin}}) = 11$  **constraints**
- $14 - 11 = 3$  **integrals** => Integrate over neutrino  $p_z$  and jet energy of both jets.

Event probability for signal and background hypothesis:

$$P(p_i^\mu, p_{j1}^\mu, p_{j2}^\mu) = \frac{1}{\sigma} \int d\rho_{j1} d\rho_{j2} dp_v^z \sum_{\text{comb}} |M(p_i^\mu)|^2 \frac{f(q_1) f(q_2)}{|q_1| |q_2|} \phi_4 W_{\text{jet}}(E_{\text{jet}}, E_{\text{part}})$$

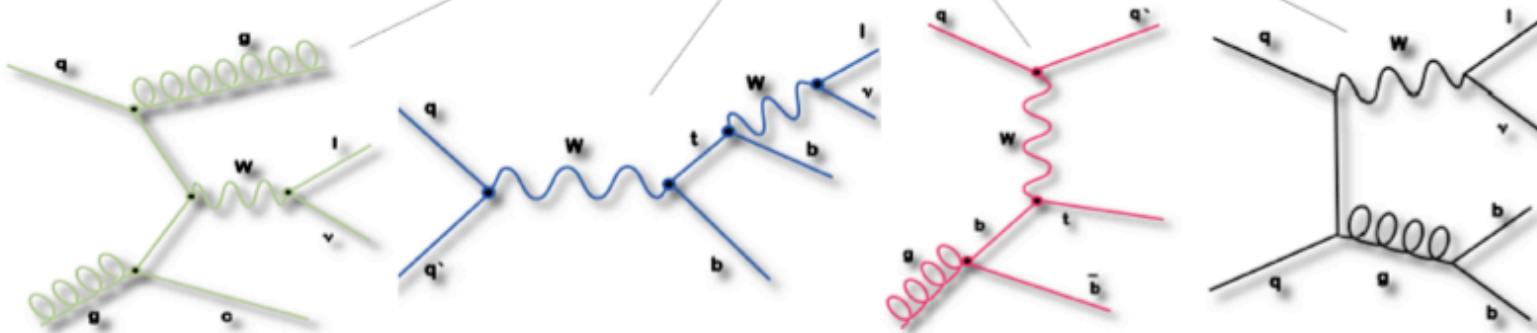
Leading Order matrix element (MadEvent)

$W(E_{\text{jet}}, E_{\text{part}})$  is the probability of measuring a jet energy  $E_{\text{jet}}$  when  $E_{\text{part}}$  was produced

Input only lepton and 2 jets 4-vectors!

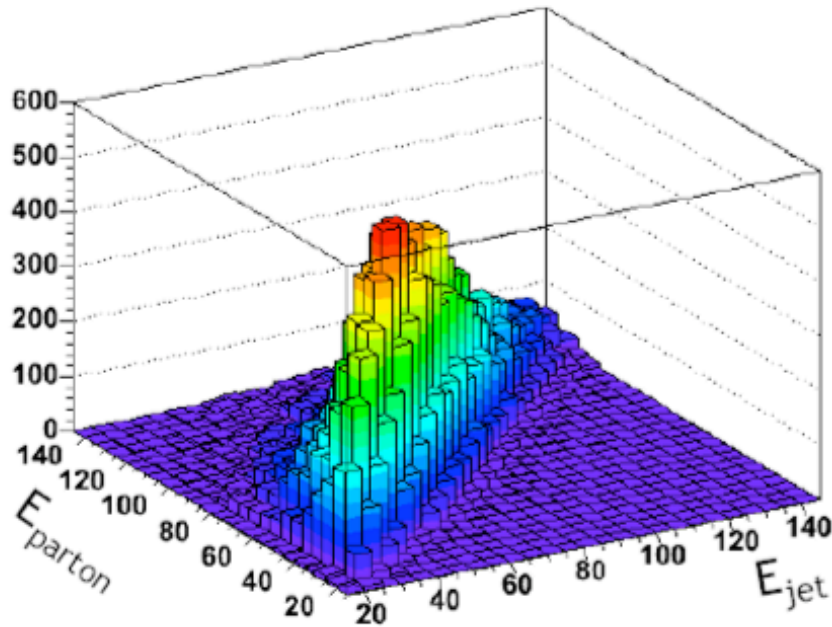
Integration over part of the phase space  $\Phi_4$

Parton distribution function (CTEQ5)

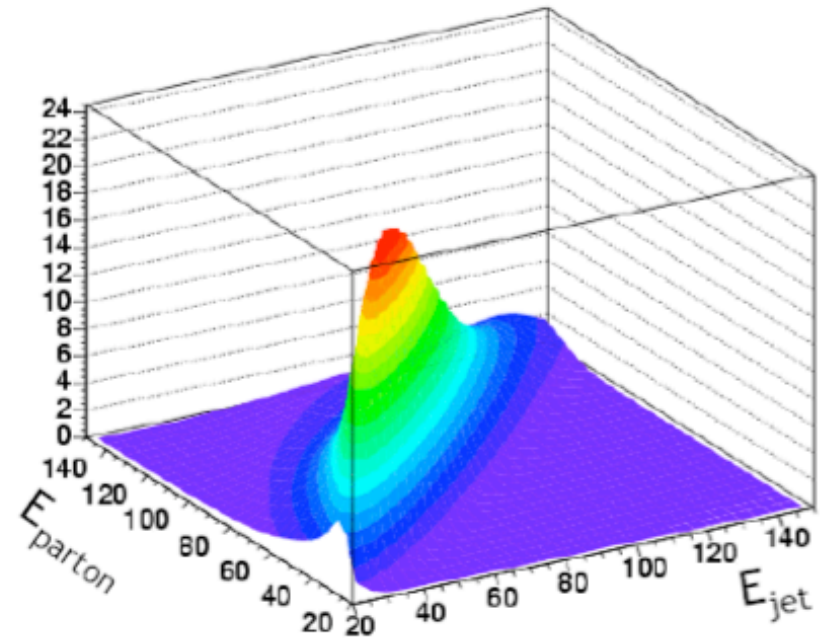


# Matrix element method – CDF transfer functions

Full simulation vs parton energy:



Double Gaussian parameterization:



Double Gaussian parameterization:

$$W_{jet}(E_{jet}, E_{parton}) = \frac{1}{\sqrt{2\pi}(p_1 + p_2 p_5)} \left[ \exp\left(-\frac{(\delta E - p_1)^2}{2p_2^2}\right) + p_3 \exp\left(-\frac{(\delta E - p_4)^2}{2p_5^2}\right) \right]$$

$$\text{where: } p_i = a_i + b_i E_{parton} \quad \delta E = (E_{parton} - E_{jet})$$

# Matrix element method – $D\bar{0}$ transfer functions

◆ To evaluate  $|M|^2$ , we must have initial/final state 4-vectors.

❖  $W(x,y)$  relates final state  $y$  to detector state  $x$

◆ Jets

❖ Assume angles well measured and Sole dependence on energy difference

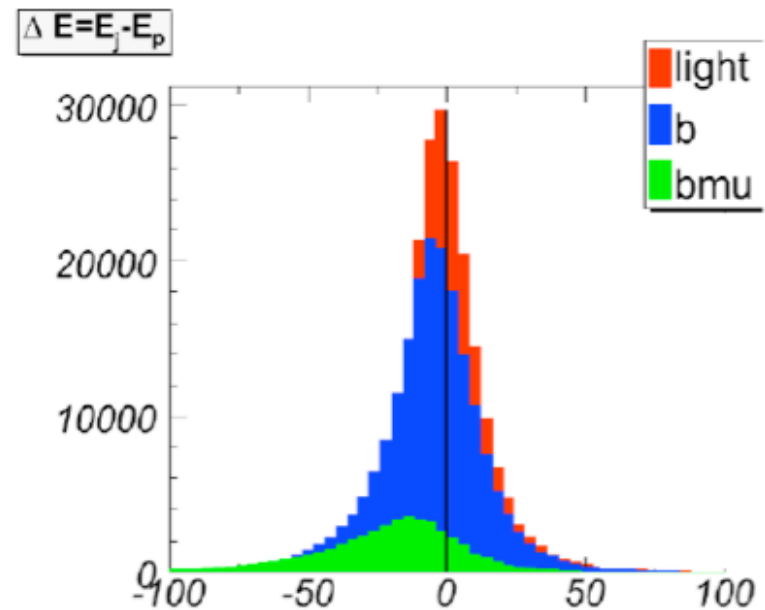
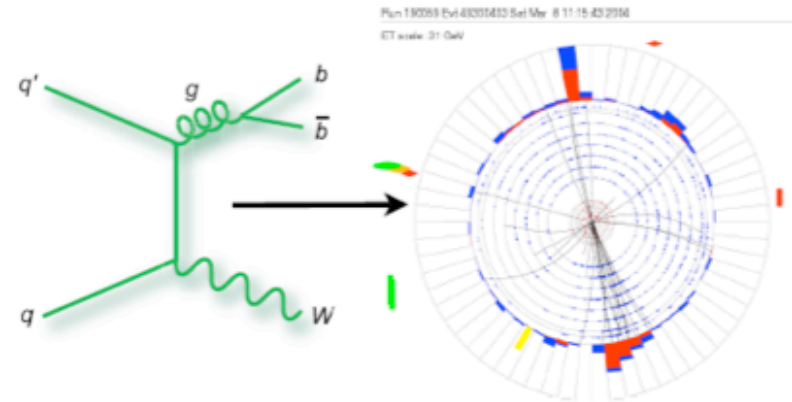
❖ Calculate for 3 types of jets: light, b, and b w/ mu

◆ Electrons

❖ Assume angles well measured and sole dependence on energy difference

◆ Muons

❖ Dependence on  $q/P_T$ ,  $\eta$ , and number of SMT hits



# W+jets heavy flavour fraction at DØ

$$\alpha(Wb\bar{b} + Wc\bar{c}) + W_{jj} + t\bar{t} + \text{QCD} = \text{Data}$$

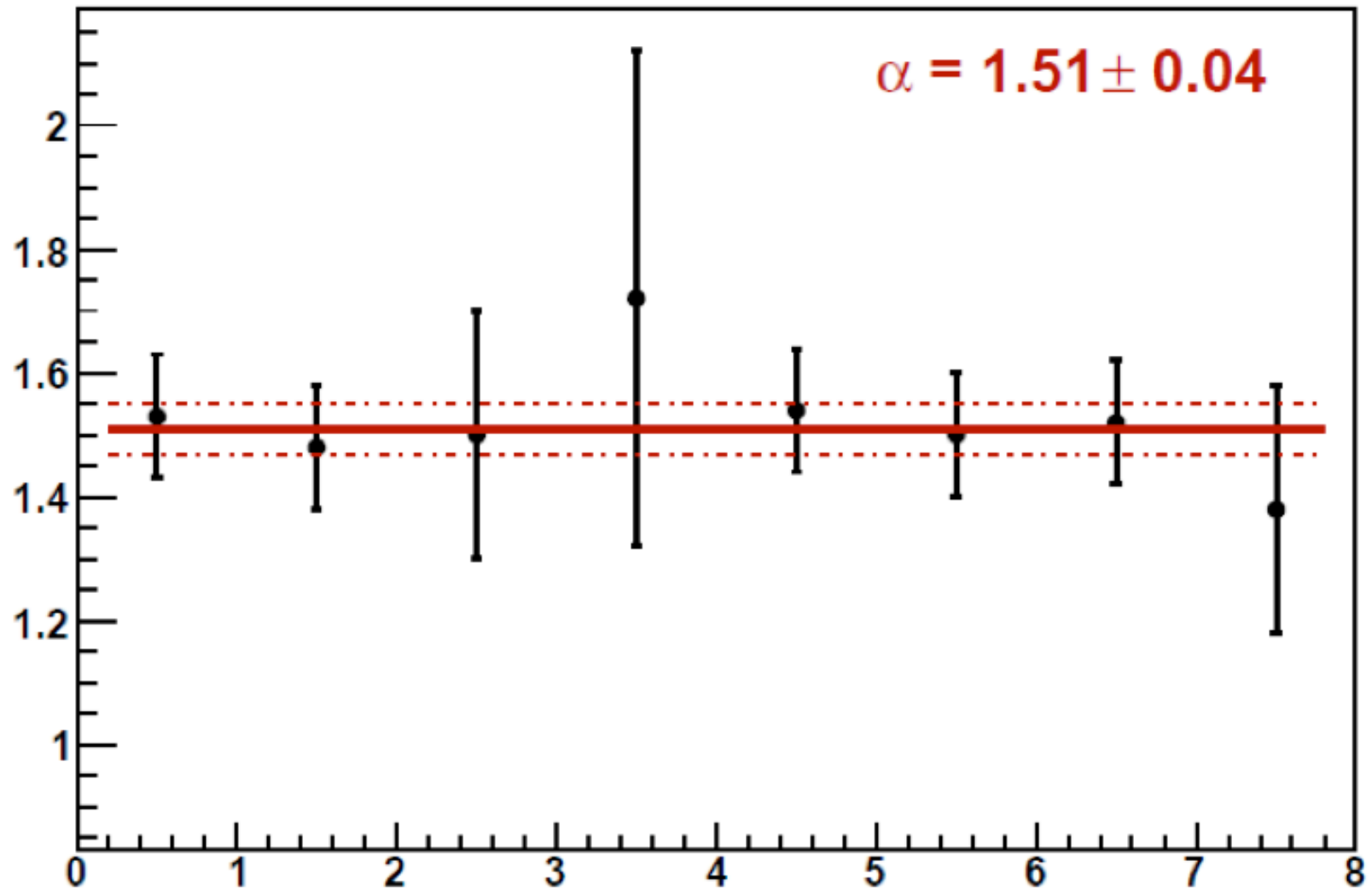
Scale Factor  $\alpha$  to Match Heavy Flavor Fraction to Data

	1 jet	2 jets	3 jets	4 jets
Electron Channel				
0 tags	$1.53 \pm 0.10$	$1.48 \pm 0.10$	$1.50 \pm 0.20$	$1.72 \pm 0.40$
1 tag	$1.29 \pm 0.10$	$1.58 \pm 0.10$	$1.40 \pm 0.20$	$0.69 \pm 0.60$
2 tags	—	$1.71 \pm 0.40$	$2.92 \pm 1.20$	$-2.91 \pm 3.50$
Muon Channel				
0 tags	$1.54 \pm 0.10$	$1.50 \pm 0.10$	$1.52 \pm 0.10$	$1.38 \pm 0.20$
1 tag	$1.11 \pm 0.10$	$1.52 \pm 0.10$	$1.32 \pm 0.20$	$1.86 \pm 0.50$
2 tags	—	$1.40 \pm 0.40$	$2.46 \pm 0.90$	$3.78 \pm 2.80$



# HF fraction – DØ

Heavy flavour scale factor  $\alpha$  measured in the zero tag bins



# HF Fraction – CDF (B. Stelzer)

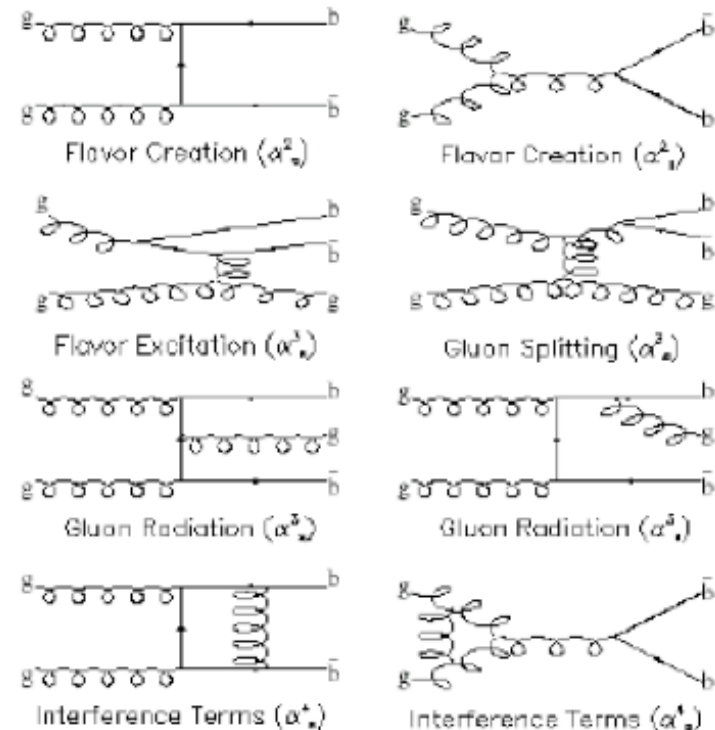
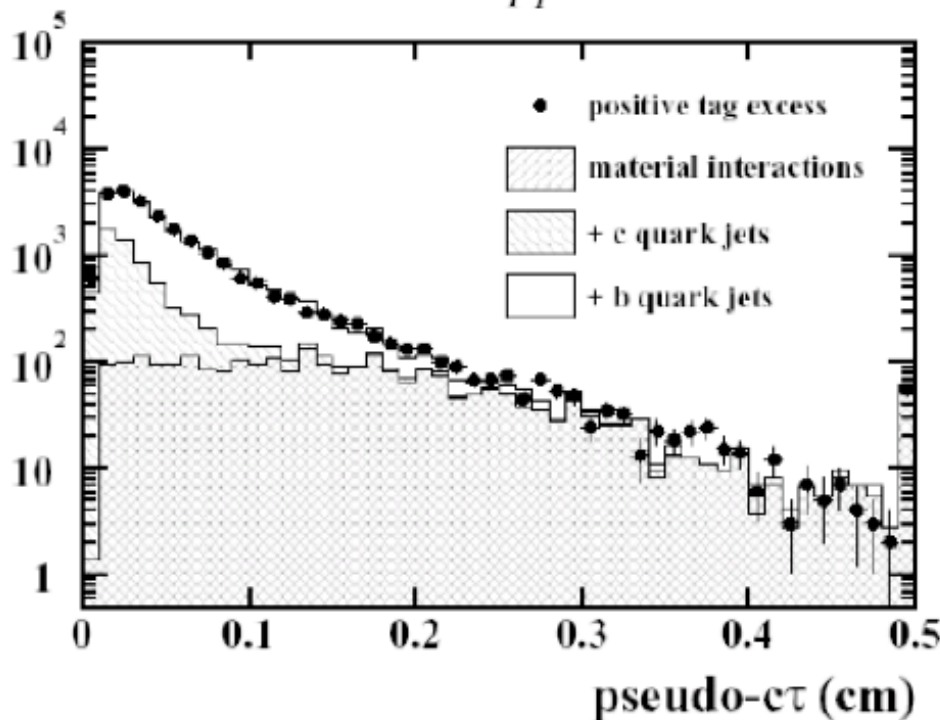
- 1) Estimate generic jet heavy flavor fraction in ALPGEN Monte Carlo
- 2) Fit for bottom and charm fraction in generic jet data

Difference between the two outcomes suggests  $K=1.5 \pm 0.4$

Result supported by study using MCFM:

J. M. Campbell, J. Houston,  
Method 2 at NLO, hep-ph/0405276

$$pseudo-c\tau = L_{2D} \cdot \frac{M^{vtx}}{p_T^{vtx}}$$



# Binned likelihood fit at CDF (B. Stelzer)

Binned Likelihood Function:

$$\mathcal{L}(\beta_1, \dots, \beta_5; \delta_1, \dots, \delta_{10}) = \underbrace{\prod_{k=1}^B \frac{e^{-\mu_k} \cdot \mu_k^{n_k}}{n_k!}}_{\text{Poisson term}} \cdot \underbrace{\prod_{j=2}^5 G(\beta_j | 1, \Delta_j)}_{\text{Gauss constraints}} \cdot \underbrace{\prod_{i=1}^{10} G(\delta_i, 0, 1)}_{\text{Systematics}}$$

Expected mean in bin k:

$$\mu_k = \sum_{j=1}^5 \beta_j \cdot \underbrace{\left\{ \prod_{i=1}^{10} [1 + |\delta_i| \cdot (\epsilon_{ji+} H(\delta_i) + \epsilon_{ji-} H(-\delta_i))] \right\}}_{\text{Normalization Uncertainty}}$$

$$\cdot \underbrace{\alpha_{jk}}_{\text{Shape } P.} \cdot \underbrace{\left\{ \prod_{i=1}^{10} (1 + |\delta_i| \cdot (\kappa_{jik+} H(\delta_i) + \kappa_{jik-} H(-\delta_i))) \right\}}_{\text{Shape Uncertainty}}$$

$\beta_j = \sigma_j / \sigma_{SM}$  parameter  
single top (j=1)

W+bottom (j=2)

W+charm (j=3)

Mistags (j=4)

ttbar (j=5)

k = Bin index

i = Systematic effect

$\delta_i$  = Strength of effect

$\epsilon_{ji\pm}$  =  $\pm 1\sigma$  norm. shifts

$\kappa_{jik\pm}$  =  $\pm 1\sigma$  shift in bin k

• All sources of systematic uncertainty included as nuisance parameters

• Correlation between Shape/Normalization uncertainty considered ( $\delta_i$ )

# Measuring cross sections at DØ

Probability to observe data distribution  $D$ ,  
expecting  $y$ :

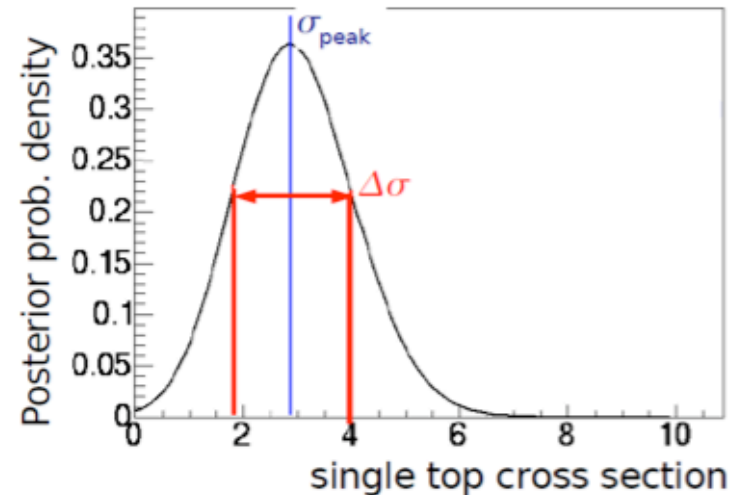
$$y = \alpha l \sigma + \sum_{s=1}^N b_s \equiv a \sigma + \sum_{s=1}^N b_s$$

$$P(D|y) \equiv P(D|\sigma, a, b) = \prod_{i=1}^{nbins} P(D_i|y_i)$$

The cross section is obtained

$$Post(\sigma|D) \equiv P(\sigma|D) \propto \int_a \int_b P(D|\sigma, a, b) Prior(\sigma) Prior(a, b)$$

- Bayesian posterior probability density
- Shape and normalization systematics treated as nuisance parameters
- Correlations between uncertainties properly accounted for
- Flat prior in signal cross section



# $|V_{tb}|$ determination

- No assumptions on the number of families or unitarity of the CKM matrix
- However, some other model assumptions have been made
- It is assumed that the only existing production mechanism of single top quarks involves the interaction with a  $W$  boson (models where single top quark events can be produced e.g. via FCNC interactions or heavy scalar or vector boson exchange, are not considered)
- Assuming  $|V_{td}|^2 + |V_{ts}|^2 \ll |V_{tb}|^2$ , implying  $B(t \rightarrow Wb) \simeq 100\%$
- Finally,  $tbW$  interaction is CP-conserving and of the  $V-A$  type, but it is allowed to have an anomalous strength
- Most general  $tbW$  vertex:

$$\Gamma_{tbW}^\mu = -\frac{g}{\sqrt{2}} V_{tb} \bar{u}(p_b) \left[ \gamma^\mu (f_1^L P_L + f_1^R P_R) - \frac{i\sigma^{\mu\nu}}{M_W} (f_2^L P_L + f_2^R P_R) \right] u(p_t)$$

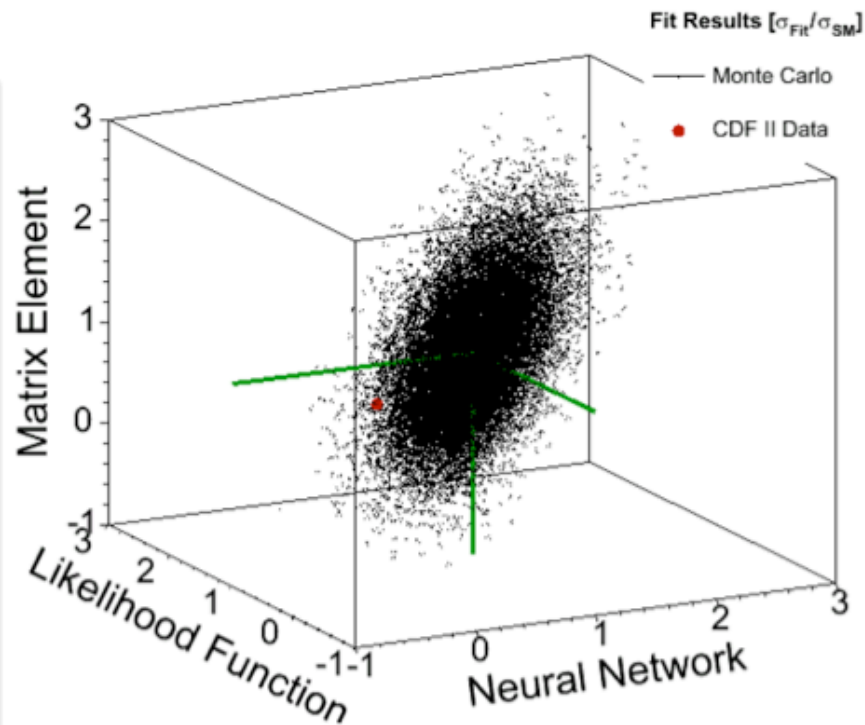
- SM: CP is conserved in the  $tbW$  vertex,  $f_1^L = 1$  and  $f_1^R = f_2^L = f_2^R = 0$

# CDF compatibility study

- ME, NN and LF analyses use same input dataset and MC events, but results differ

## Potential sources of differences

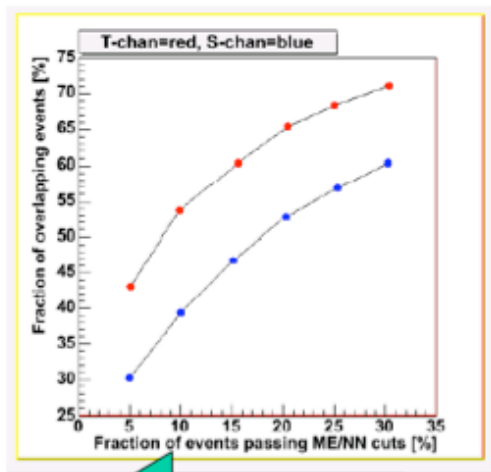
- ME uses transfer functions
- ME does not use missing ET
- ME integrates over all neutrino  $p_z$ , while NN chooses the solution with smaller  $|p_z|$
- with two jets in the event, the NN choose the secondary-vertex-tagged jet as the  $b$  jet from top quark decay. The ME sums over both possibilities
- NN also allows for soft jets ( $8 < E_T(\text{jet}) < 15 \text{ GeV}$ )



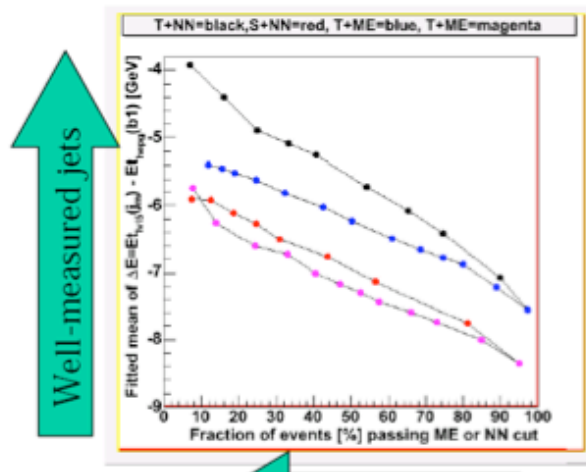
- LF(t), ME(s+t), NN(s+t)
- Coming: all six discriminants: LF(s), LF(t), ME(s+t), NN(s), NN(t), NN(s+t)

# CDF compatibility study (cont'd)

- Overlap between 5% highest-ME and 5% highest-NN(s+t) events is 30(43)% for s(t)-channel (left plot)
- Impact of transfer functions (middle plot): NN needs better-measured jets in signal region (close to 0) than ME. Significant effect in t-channel only (black/blue curves)
- Missing ET measurement (right plot): NN needs better-measured MET in signal region (close to 0) than ME

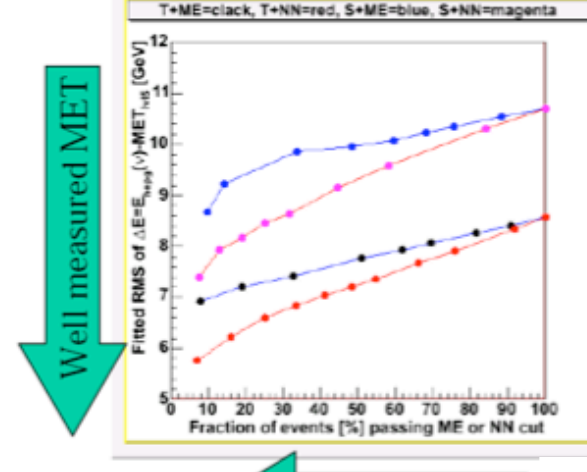


← Signal-likeness



↑ Well-measured jets

← Signal-likeness



↓ Well measured MET

← Signal-likeness