



2007 Europhysics Conference in Manchester, England

UNIVERSITY OF
CALIFORNIA
Riverside

Evidence for Single Top Quark Production at DØ

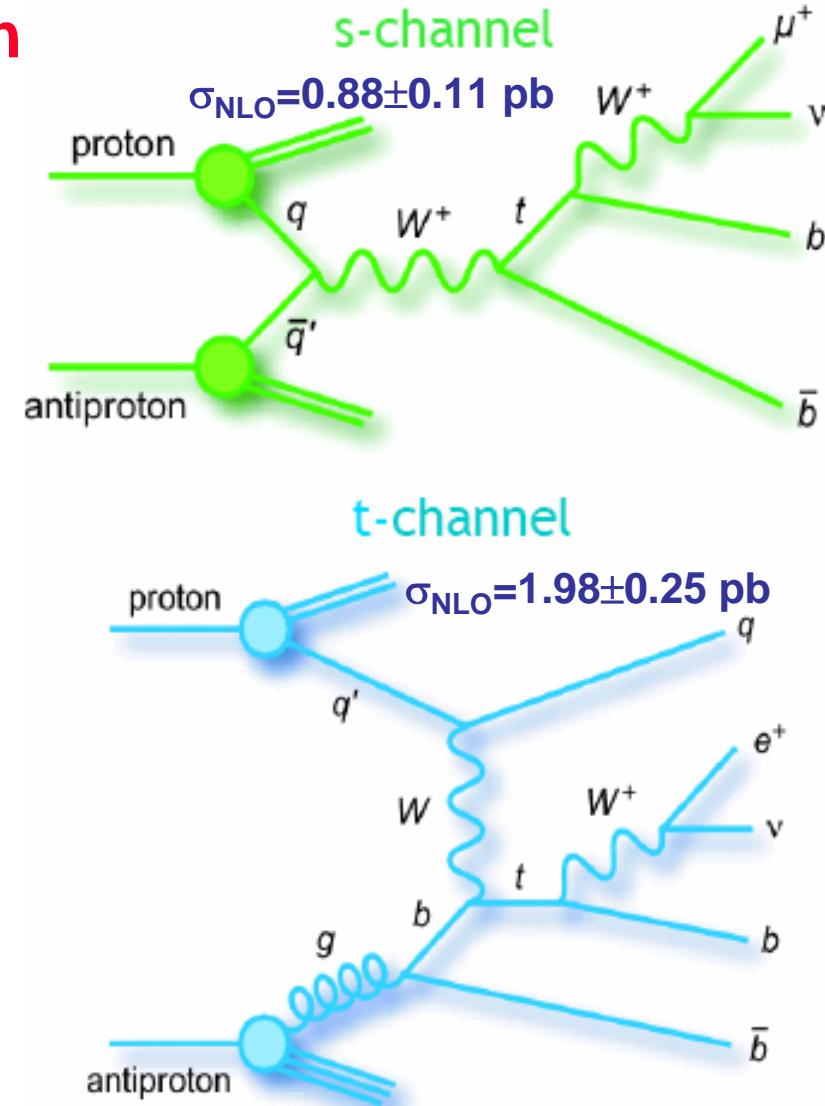
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On Behalf of the DØ Collaboration

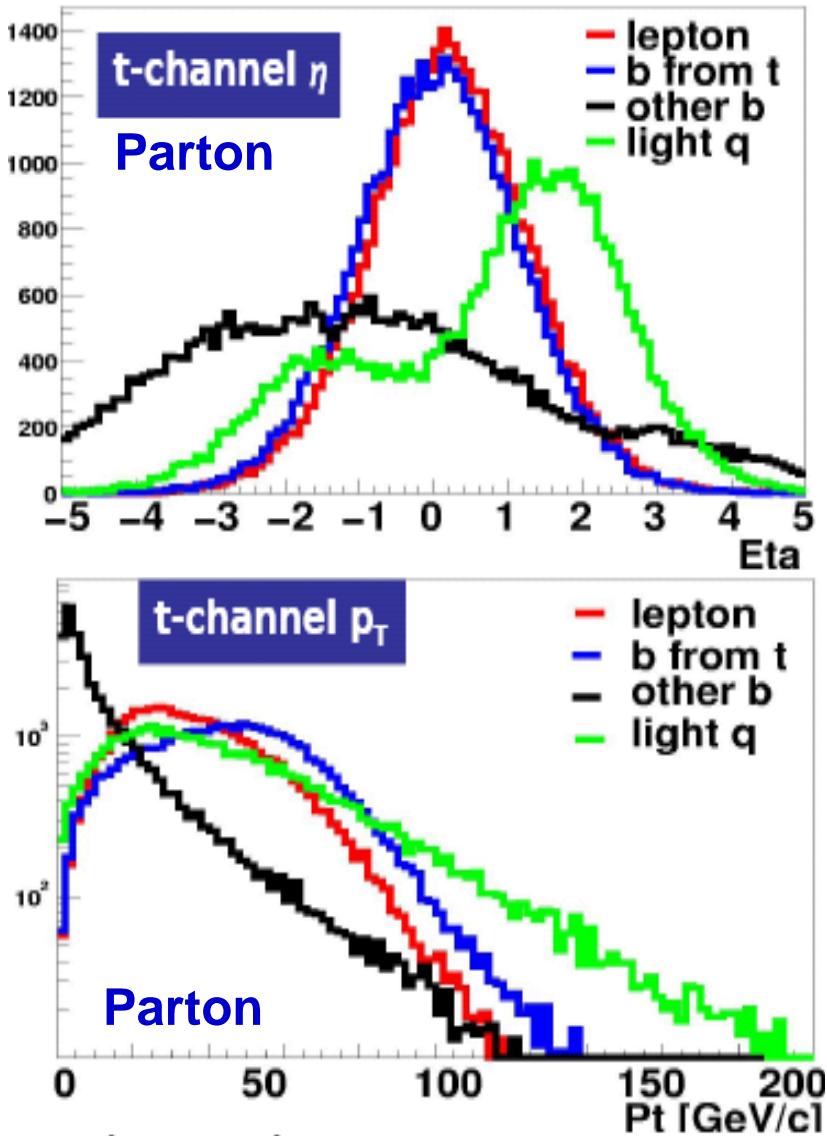
Single Top Quark Production

Single Top: Electroweak Interaction

- Study Wtb coupling
 - Direct measurement of the $|V_{tb}|$ CKM matrix element
 - Test of CKM unitarity
 - Anomalous Wtb couplings
- New physics, example:
 - s-channel is sensitive to W' , H^+
 - t-channel is sensitive to FCNC
 - 4th quark generation?
- Study top properties:
 - Polarization, decay width, lifetime...
- Background study helps many physics searches, e.g. SM Higgs



Event Selection



Event Topology:

- Energetic isolated lepton (from W)
- Missing E_T (ν from W)
- One b-quark jet (from top)
- A light flavor jet and/or another b-jet

One isolated lepton

- Electron $p_T > 15 \text{ GeV}$, $|\eta| < 1.1$
- Muon $p_T > 18 \text{ GeV}$, $|\eta| < 2.0$

Missing $E_T > 15 \text{ GeV}$

One or two b-tagged jet

- 2–4 jets: $p_T > 15 \text{ GeV}$, $|\eta| < 3.4$
- Leading jet: $p_T > 25 \text{ GeV}$, $|\eta| < 2.5$
- Second leading jet: $p_T > 20 \text{ GeV}$

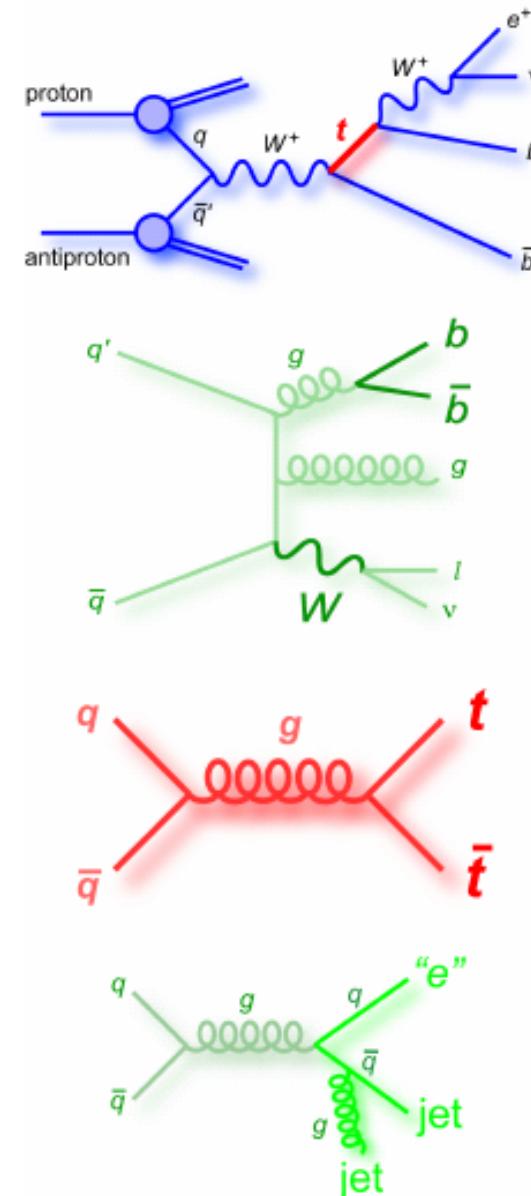
Signal and Background Modeling

Signal:

- CompHEP-SINGLETOP
- Distributions agree well with ZTOP & MCFM (NLO)

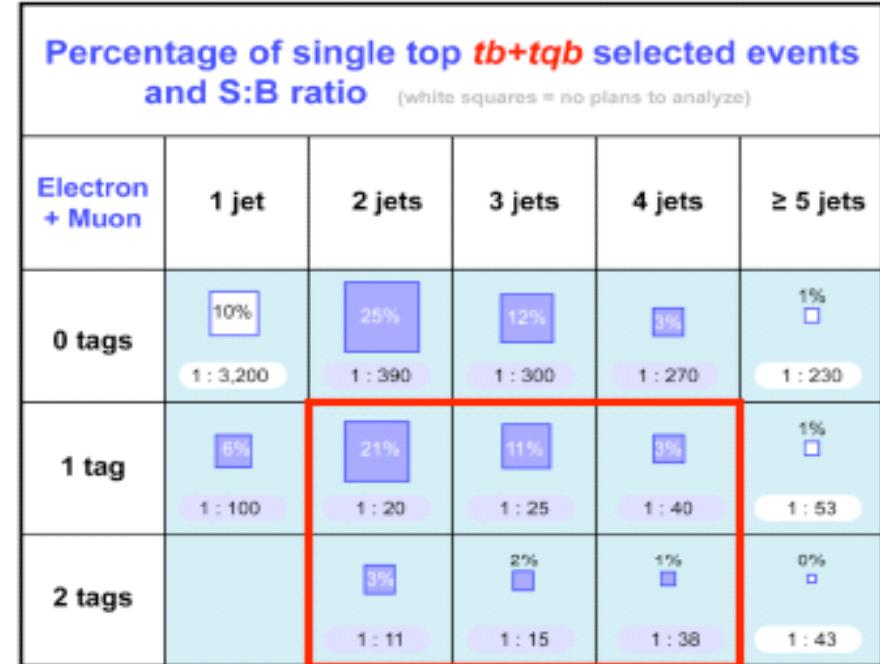
Background:

- W+jets production
 - Estimated from data & MC
 - Distribution shapes from ALPGEN
 - Normalization, Wcc and Wbb factor from data
- Top pair production
 - ALPGEN
 - Normalized to NNLO cross section
- Multijet events
 - Misidentified lepton
 - Estimated from data



Event Yields and Systematics

Source	Event Yields in 0.9 fb^{-1} Data Electron+muon, 1tag+2tags combined		
	2 jets	3 jets	4 jets
$t\bar{b}$	16 ± 3	8 ± 2	2 ± 1
tqb	20 ± 4	12 ± 3	4 ± 1
$t\bar{t} \rightarrow ll$	39 ± 9	32 ± 7	11 ± 3
$t\bar{t} \rightarrow l+jets$	20 ± 5	103 ± 25	143 ± 33
$W+b\bar{b}$	261 ± 55	120 ± 24	35 ± 7
$W+c\bar{c}$	151 ± 31	85 ± 17	23 ± 5
$W+jj$	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



Component	Size
W+jets&QCD normalization	18 – 28%
top pair normalization	18%
Tag rate functions (shape)	2 – 16%
Jet energy scale (shape)	1 – 20%
Luminosity	6%
Trigger modeling	3 – 6%
Lepton ID	2 – 7%
Jet modeling	2 – 7%
Other small components	few%

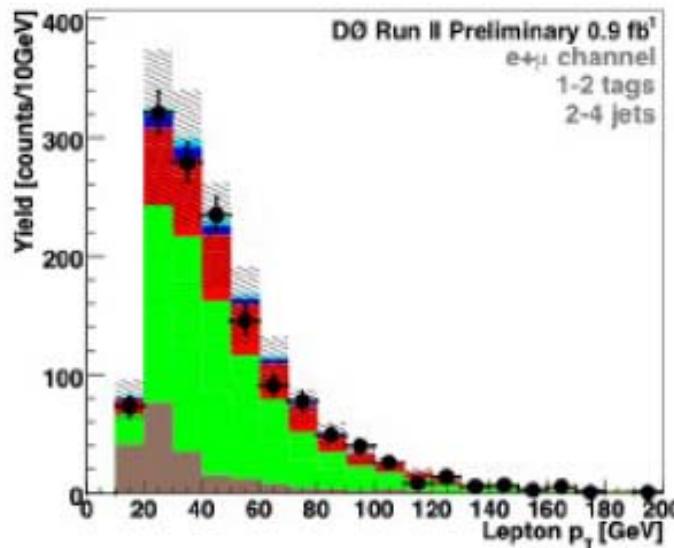
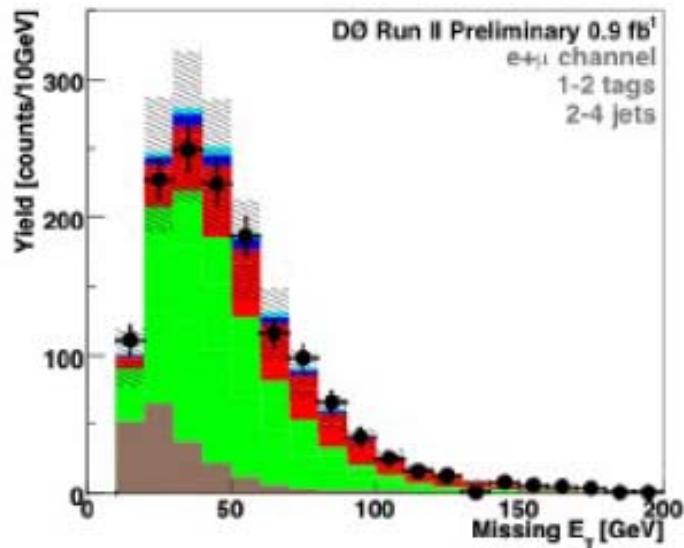
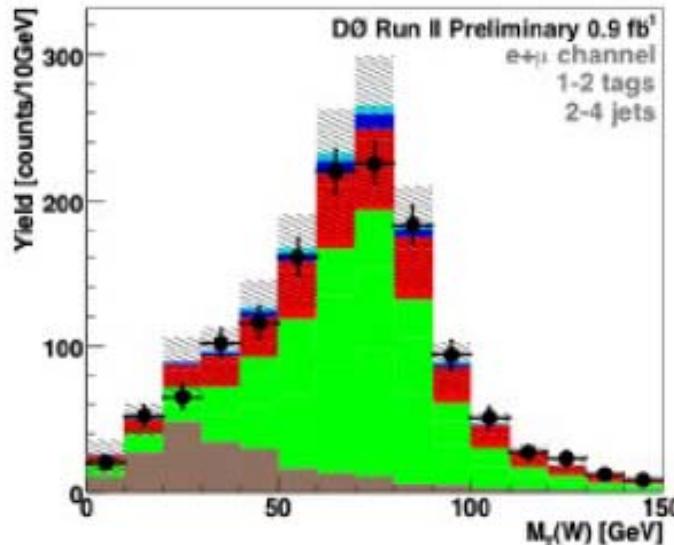
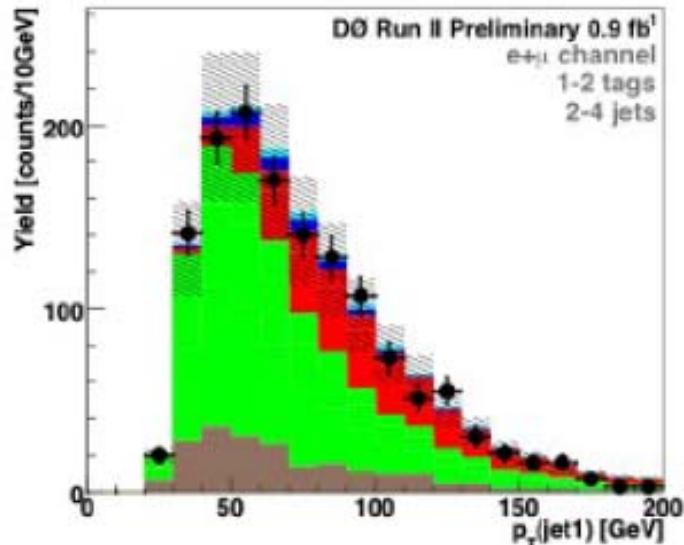
Expect 62 signal and 1398 bkgd events

Use multivariate discriminant to separate signal from background

Most systematic uncertainties apply only to normalization, except jet energy scale and b-tagging which affect shapes

Cross section uncertainties are dominated by the statistical uncertainty

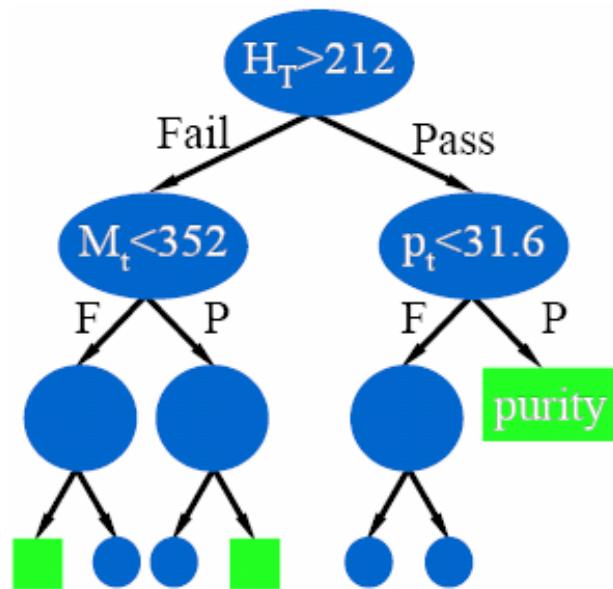
Background Modeling



Key for Plots

- Data
- tb
- tqb
- tt̄
- W + jets
- Multijets
- ±1 σ uncertainty on background

Decision Tree Method



Train on 1/3 of bkgd and signal sample

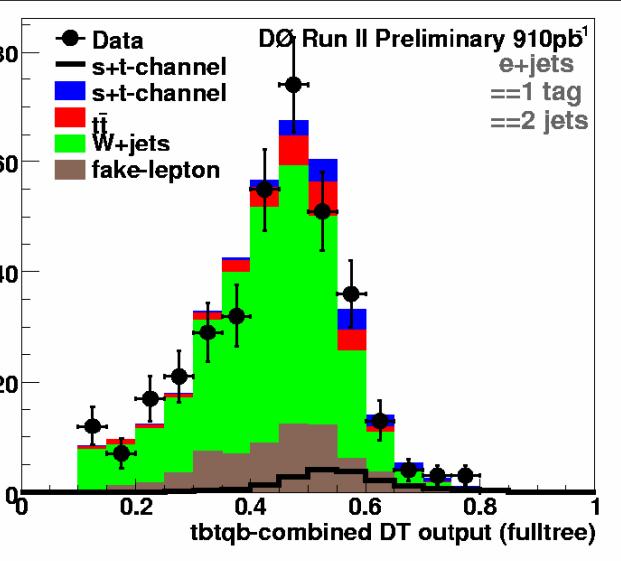
- Creates a tree, with a simple straight cut at every branch point
- Select variable and splitting value with best separation, repeat recursively. Stop if improvement stops or too few events
- Each leaf classifies an event with a purity $N_s/(N_s+N_B)$
- Measurements done on the other 2/3's of signal and background sample

Boosting:

- Retrain 20 times to learn from misclassified events

Discriminating variables:

- 49 physics motivated variables
- $M_{\text{all jets}}$, $M_{W,b\text{-jet}1}$, $\cos_{b\text{-jet}1,\text{lepton}}$, $q_{\text{lepton}} * \eta_{\text{light-jet}}$



Matrix Element Method

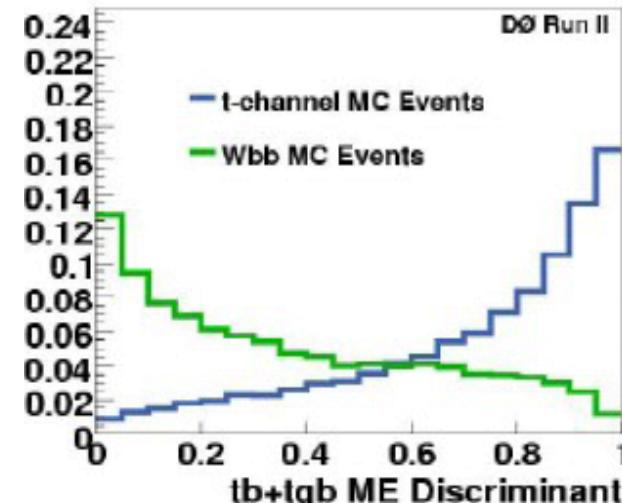
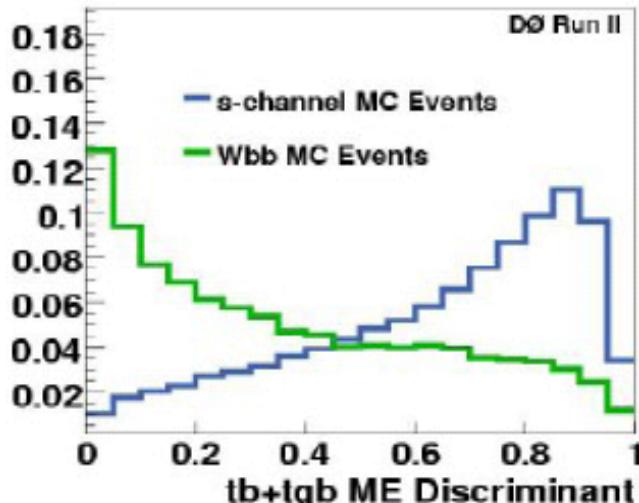
Use full kinematical information contained in event: the four vectors from the reconstructed lepton and jets

Use matrix elements of main signal and background
Feynman diagrams to compute an event probability density
for signal and background hypotheses

$$P(\vec{x}) = \frac{1}{\sigma} \int f(q_1; Q) dq_1 f(q_2; Q) dq_2 \times |M(\vec{y})|^2 \phi(\vec{y}) dy \times W(\vec{x}, \vec{y})$$

Parton distribution functions CTEQ6 Differential cross section (LO ME from Madgraph) Transfer Function: maps parton level (y) to reconstructed variables (x)

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

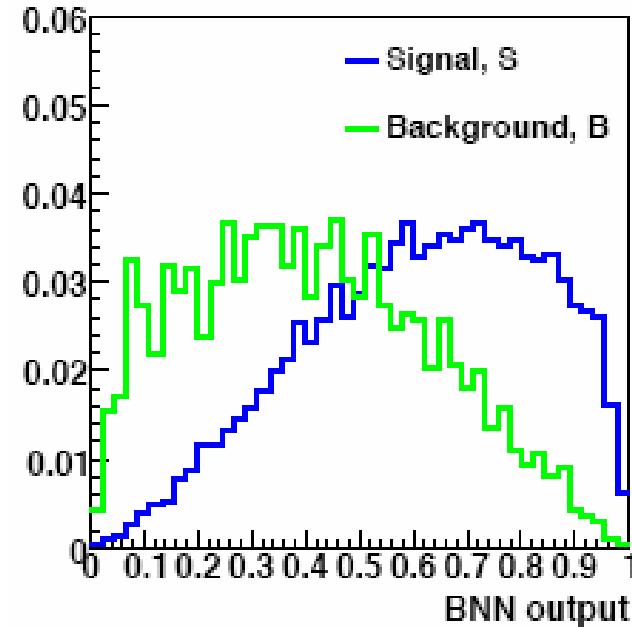
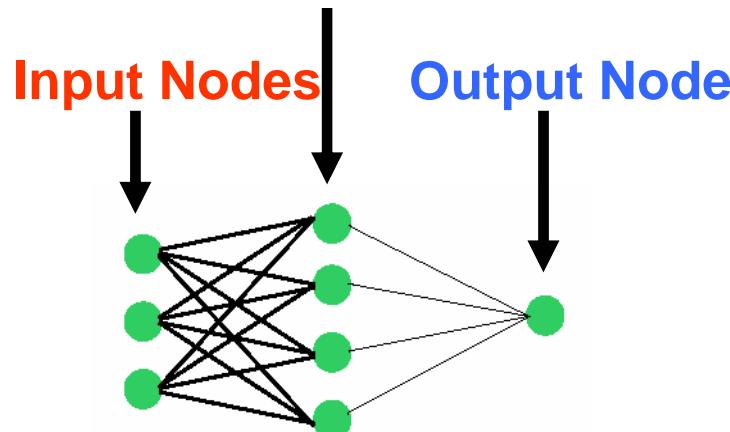


Bayesian Neural Networks Method

Neural networks use many input variables, train on signal and background samples, produce one output discriminant

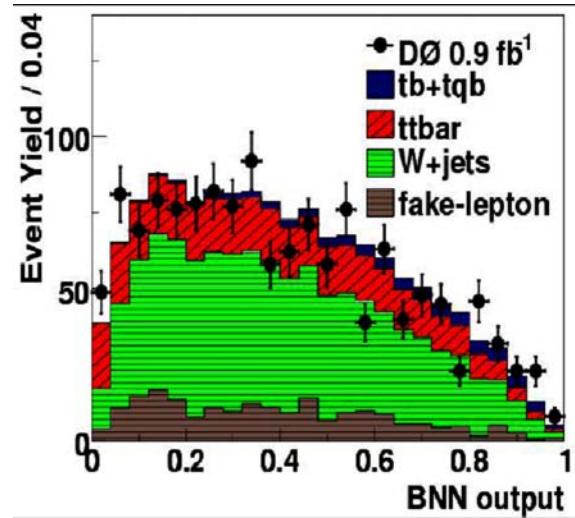
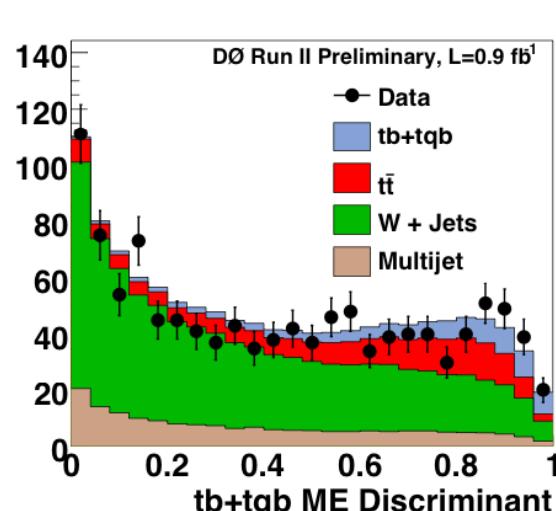
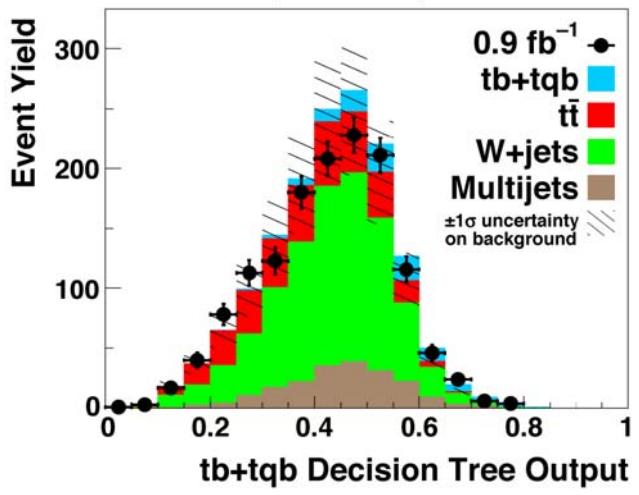
Bayesian neural networks improve on this technique:

- Average over many networks weighted by the probability of each network given the training samples
- Less prone to over-training
- Network structure is less important
 - Can use larger numbers of variables and hidden nodes
 - 20 input variables (subset of DT)
 - 20 hidden nodes

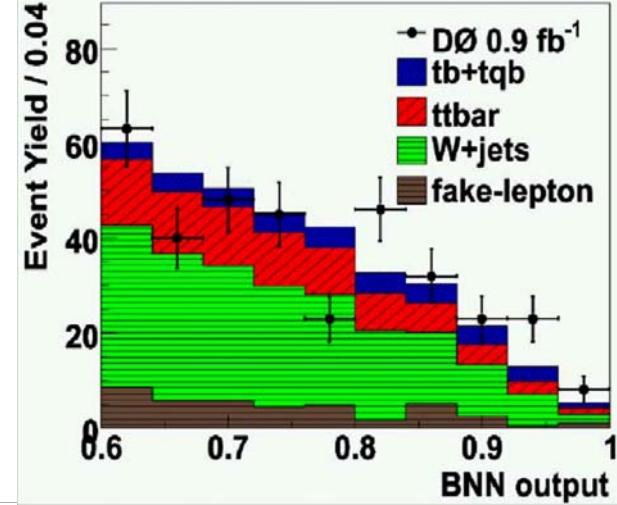
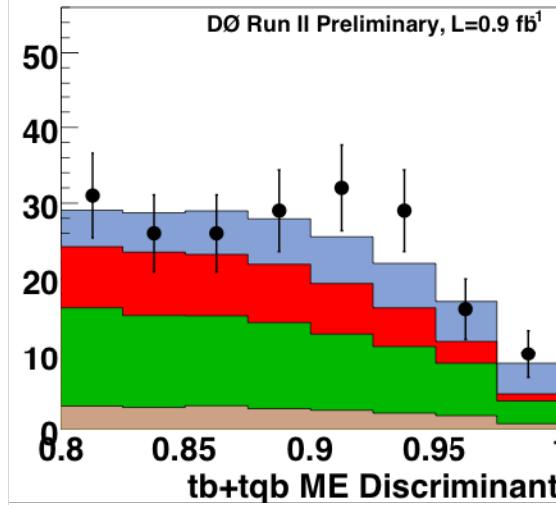
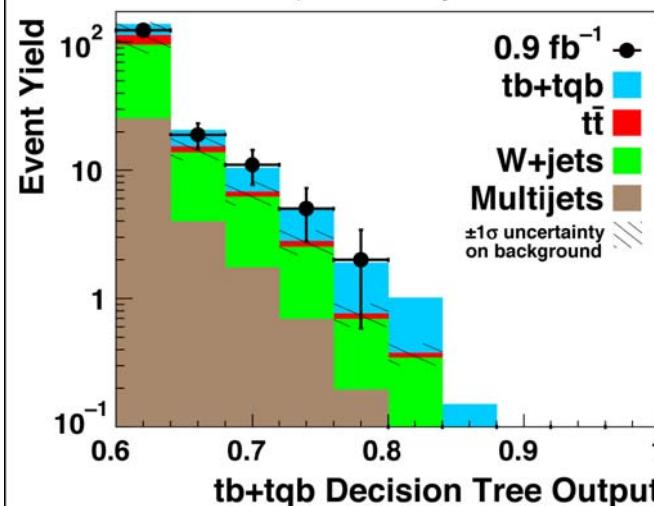


Discriminant Output

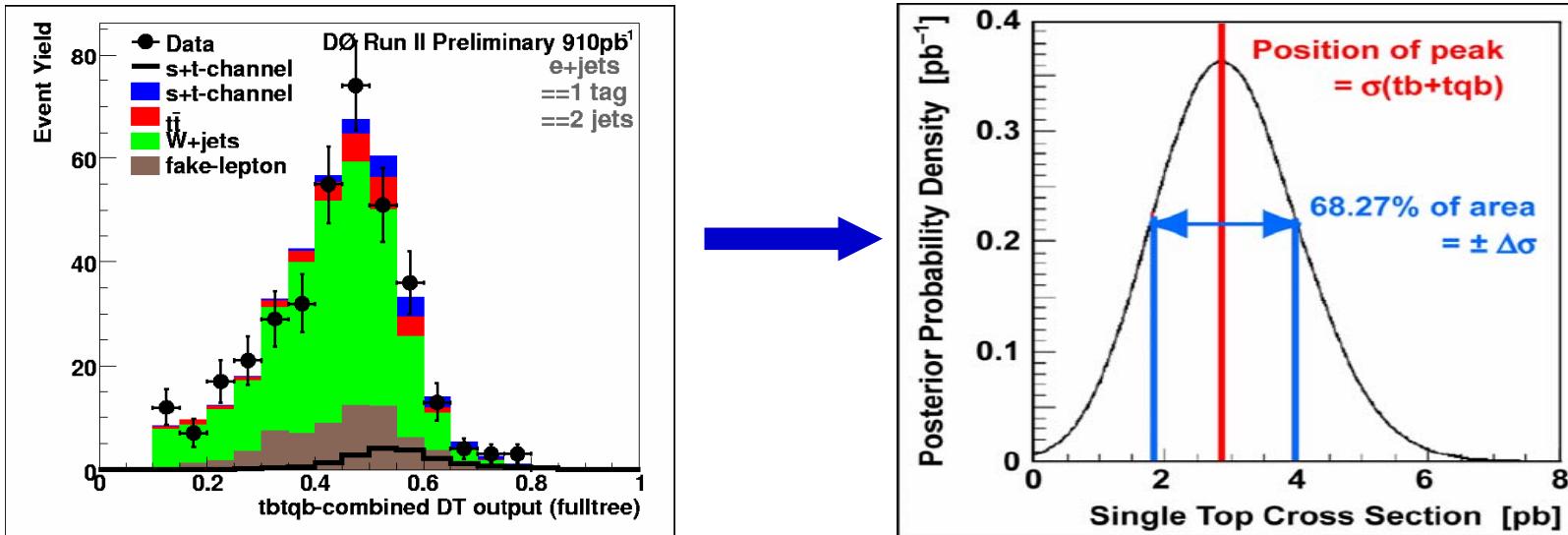
DØ Run II preliminary



DØ Run II preliminary



Cross Section Measurement



Binned likelihood from discriminant distribution

Compute Bayesian posterior probability density as a function of $\sigma(t\bar{b} + t\bar{q}b)$

- Flat prior for the cross section
- Systematic uncertainties are treated as nuisance parameters

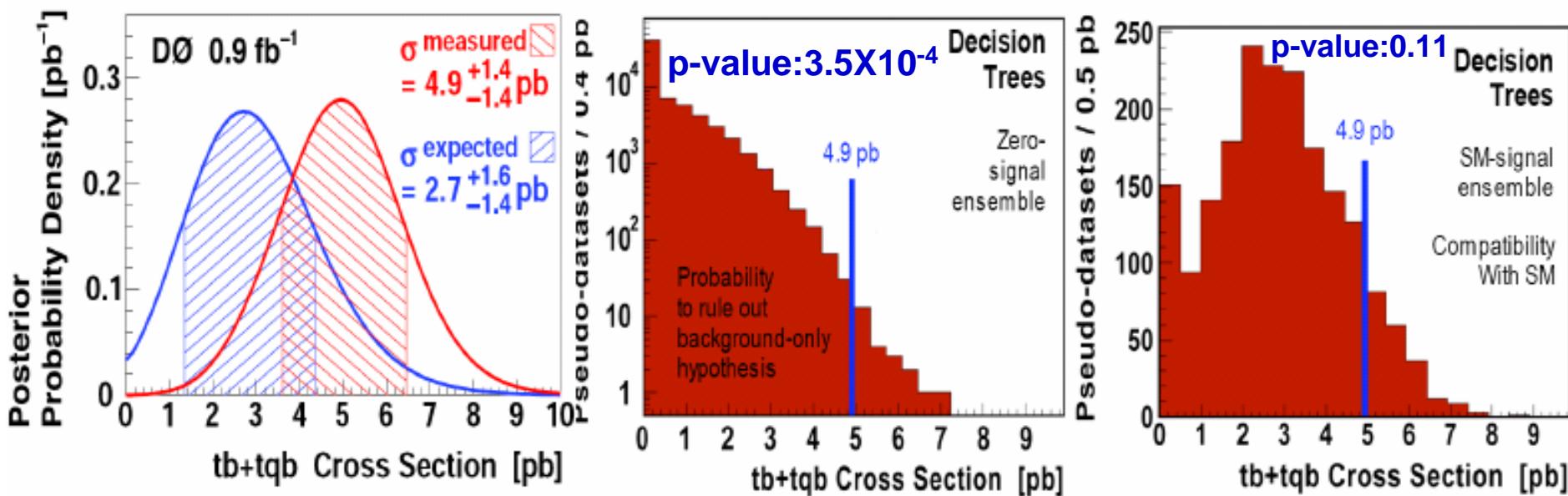
Significance as in “excess in data over background”

- P-value: assuming a null hypothesis, what’s the probability to get a value equal to or greater than the value observed
- We measure the fraction of zero-signal datasets in which we derive at least the SM cross section (**expected significance**), or at least the observed cross section (**observed significance**)

Expected and Observed Results

	Bayesian NN		Matrix Element		Decision Trees	
	Exp.	Obs.	Exp.	Obs.	Exp.	Obs.
$\sigma(tb+tqb)[pb]$	2.7 ± 1.4	$4.4^{+1.6}_{-1.4}$	$2.8^{+1.6}_{-1.4}$	$4.8^{+1.6}_{-1.4}$	$2.7^{+1.5}_{-1.4}$	4.9 ± 1.4
Significance	2.2σ	3.1σ	2.1σ	3.2σ	2.1σ	3.4σ

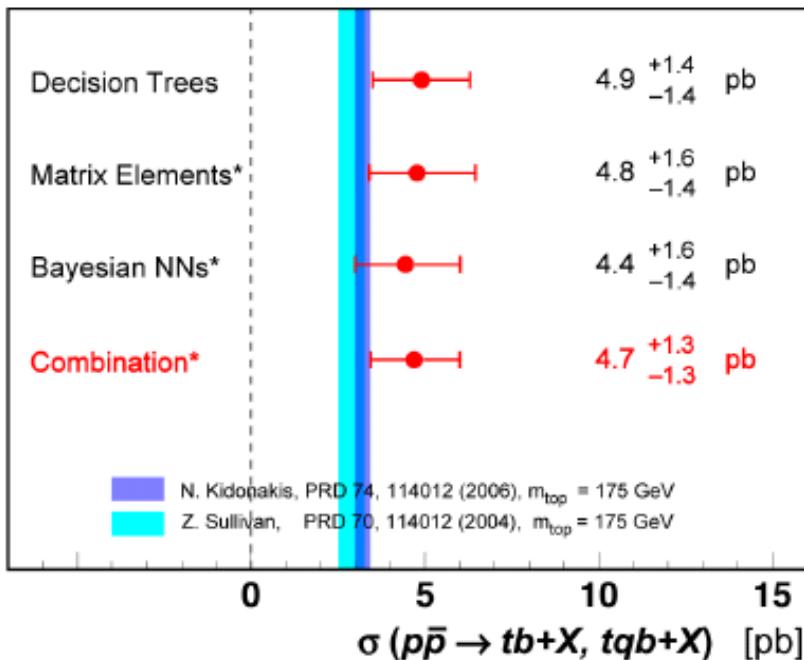
All three analyses show $> 3.0 \sigma$ excess,
Evidence for single top quark production!
SM compatibility is 11%



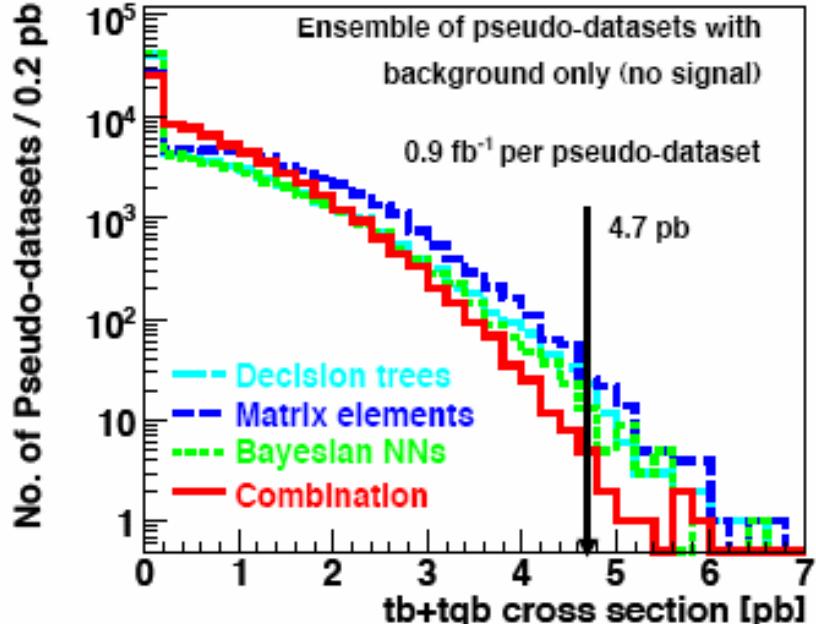
Combination for All Three Analyses

DØ Run II * = preliminary

0.9 fb^{-1}



DØ Run II Preliminary



Three analyses give consistent results

- Using same data set, thus highly correlated

Combined result using BLUE method

- 4.7 ± 1.3 pb with 3.6σ significance!

Direct measurement of $|V_{tb}|$

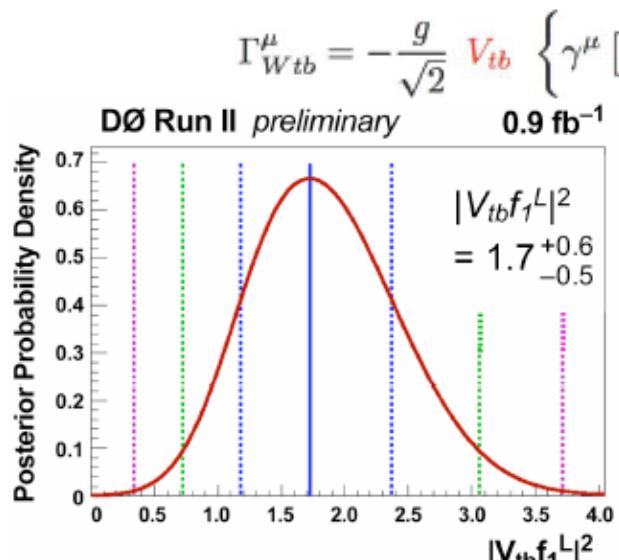
Once we have a cross section measurement, we can make the first direct measurement of $|V_{tb}|$

- Calculate posterior in $|V_{tb}|^2$: $\sigma \propto |V_{tb}|^2$

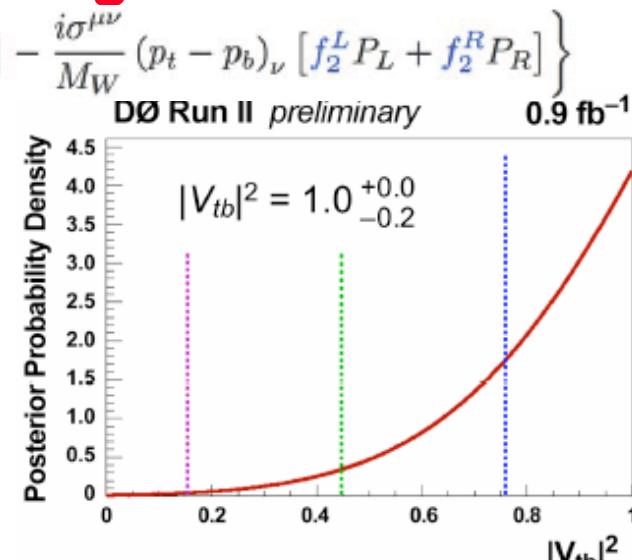
Assuming standard model production:

- Pure V-A and CP conserving interaction: $f_1^R = f_2^L = f_2^R = 0$
- $|V_{td}|^2 + |V_{ts}|^2 \ll |V_{tb}|^2$
- Additional theoretical errors needed (top mass, scale, PDF etc...)

Measurement does not assume 3 generations or unitarity



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



$$0.68 < |V_{tb}| < 1 \text{ at 95%CL (assuming } f_1^L = 1\text{)}$$

Summary

Evidence for single top production found at D0!

$\sigma(s+t) = 4.9 \pm 1.4 \text{ pb}$ with 3.4σ significance!

Analysis published: PRL 98, 181802

Analysis webpage

First direct measurement of $|V_{tb}|$

$|V_{tb}| > 0.68$ @ 95% C.L. (assuming $f_1^L = 1$)

Latest combined result: DT + ME + BNN

$\sigma(s+t) = 4.7 \pm 1.3 \text{ pb}$ with 3.6σ significance!

Outlook

Collected more than twice data used for this analysis

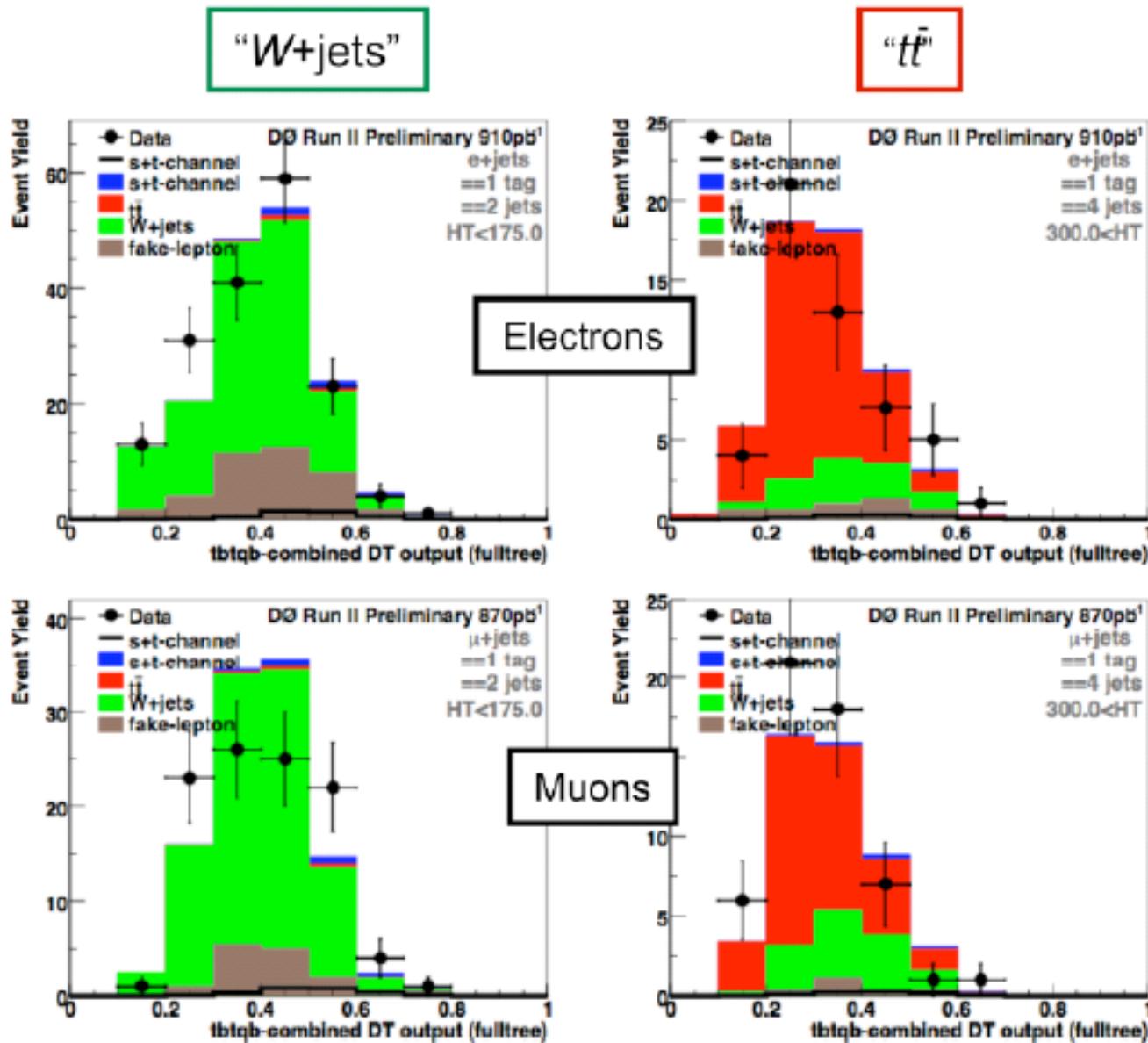
- Data delivered: $> 3 \text{ fb}^{-1}$ (goal of RunII is $4-9 \text{ fb}^{-1}$)

Expand to searches of new phenomena

- H^+ , anomalous Wtb coupling and more...

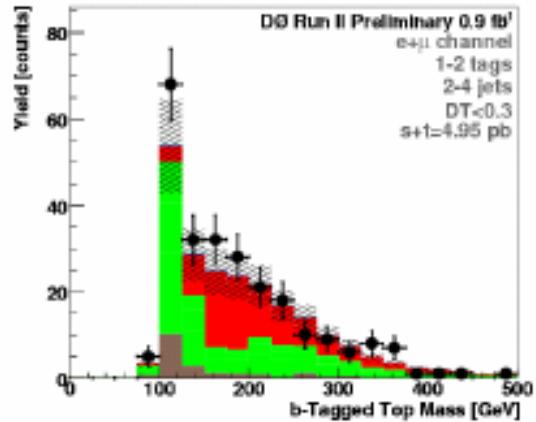
Backup Slides

Decision Tree: Cross-check Samples

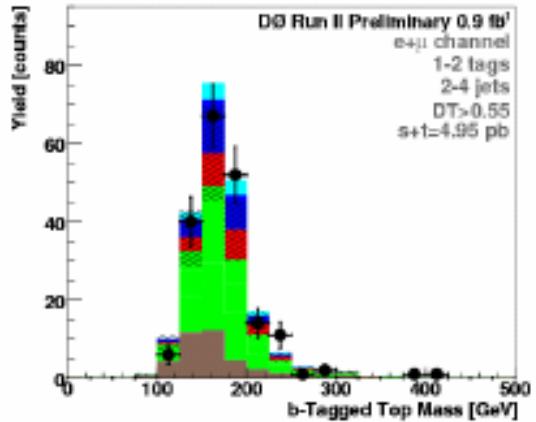


Decision Tree: Event Characteristics

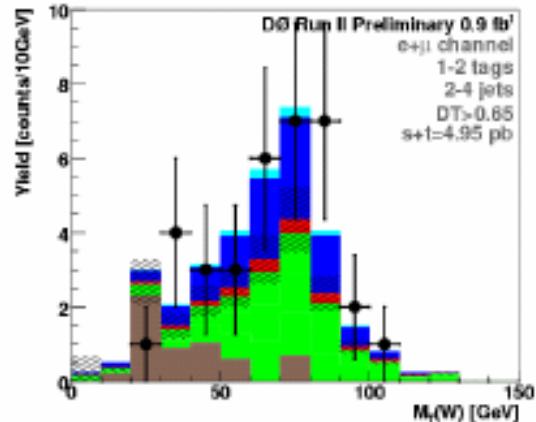
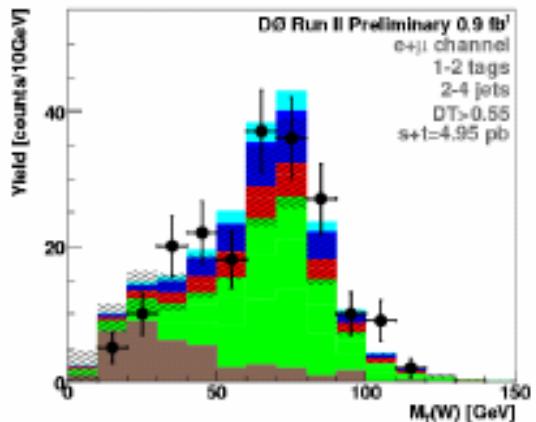
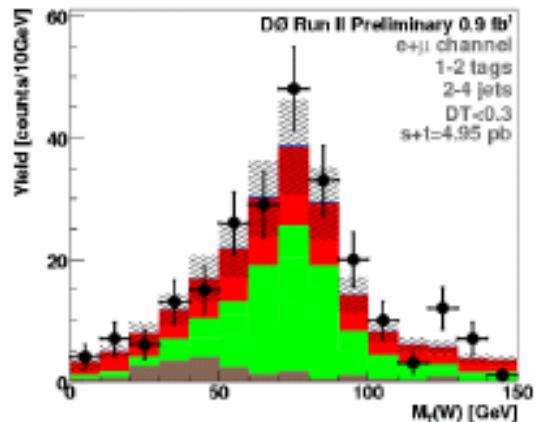
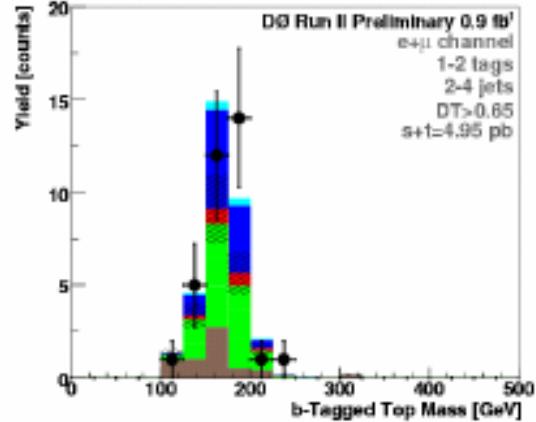
$DT < 0.3$



$DT > 0.55$



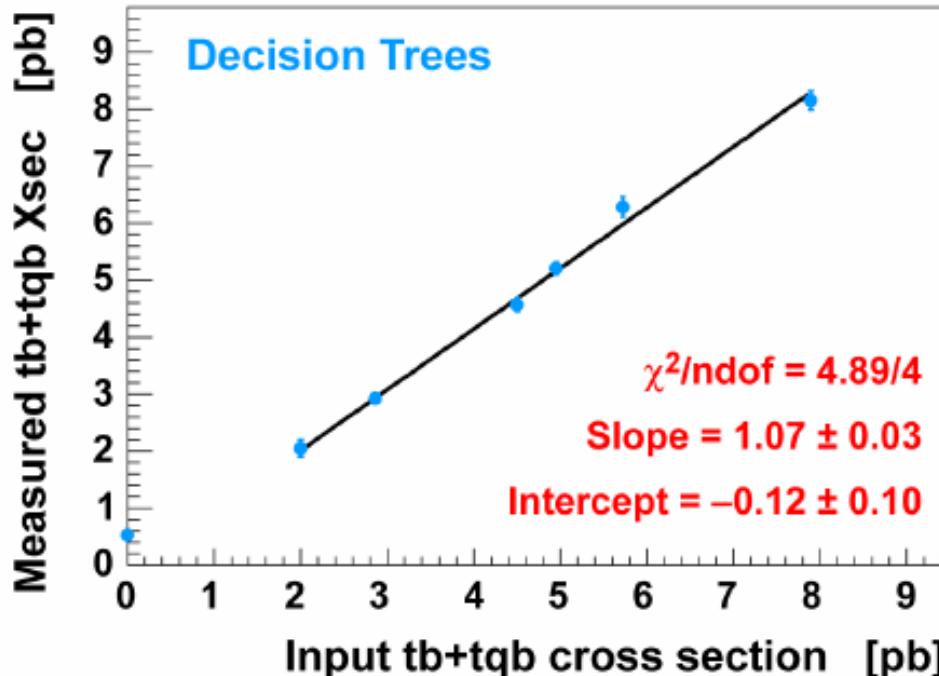
$DT > 0.65$



Decision Tree: Ensemble Testing

Tested our machinery with many sets of pseudo-data

- Subset of our total pool of background events
- Individual statistical and systematical fluctuations
- Wonderful tool – like running D0 1000s of times!
- Generated several ensembles with different single top content
- Compare measured cross sections to input ones
- Linear response, negligible bias



NN b-jet Tagger

NN trained on 7 input variables
from SVT, JLIP and CSIP taggers
Much improved performance!

- Fake rate reduced by 1/3 for same b-efficiency relative to previous tagger
- Smaller systematic uncertainty

Tag Rate Functions (TRFs) in η ,
 P_T and z-PV derived in data are
applied to MC

Our operating point:

- b-jet efficiency: ~50%
- c-jet efficiency: ~10%
- Light-jet efficiency: ~0.5%

