

Evidence for single top quark production at DØ

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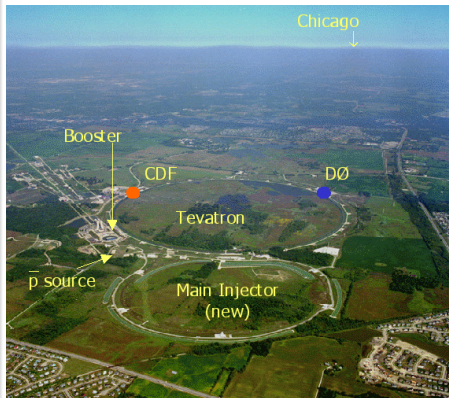


- 1 Tevatron accelerator and DØ detector
- 2 Single top quark production — Should you care?
- 3 Preparing for the measurement
 - Event selection
 - Signal and background samples
 - b tagging
- 4 Multivariate analysis techniques
- 5 Expected sensitivity
- 6 Cross sections and significance
- 7 First direct measurement of $|V_{tb}|$
- 8 New combination of analyses
- 9 Conclusion



The Tevatron at Fermilab

- Located outside Chicago, Illinois
- The world's highest-energy accelerator
- $p\bar{p}$ collider, centre-of-mass energy 1.96 TeV
- Run I: 1992-1996 at 1.8 TeV
- Started operating for Run II in March 2001
- Upgraded for Run II
 - 396 ns bunch spacing
 - new Main Injector and Recycler
 - ⇒ increased antiproton intensity

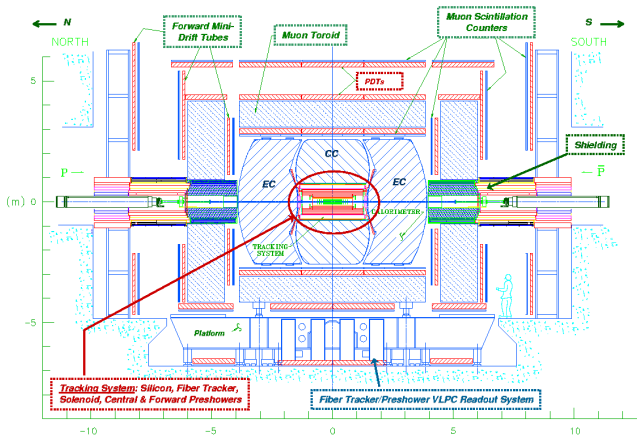


- Peak luminosity
 $> 2.5 \cdot 10^{32} \text{ cm}^{-2}\text{s}^{-1}$



The DØ detector upgrade

- 2 T superconducting solenoid
- silicon detector
- fiber tracker
- preshower detector
- upgraded muon system
- new calorimeter electronics
- upgraded trigger and DAQ



The collaboration

- 600+ physicists, 89 institutes, 18 countries



AZ U. of Arizona
CA U. of California, Berkeley
U. of California, Riverside
Cal. State U., Fresno
Lawrence Berkeley Nat. Lab.
FL Florida State U.
IL Fermilab
U. of Illinois, Chicago
Northern Illinois U.
Northwestern U.
IN Indiana U.
U. of Notre Dame
Purdue U. Calumet
IA Iowa State U.
KS U. of Kansas
Kansas State U.
LA Louisiana Tech U.
MD U. of Maryland
MA Boston U.
Northeastern U.
MI U. of Michigan
Michigan State U.
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SUNY, Stony Brook
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OK Langston U.
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LPC, Clermont-Ferrand
ISN, IN2P3, Grenoble
CPRM, IN2P3, Marseille
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U. of Aachen
Bonn U.
U. of Freiburg
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Ludwig-Maximilians U., Munich
U. of Wuppertal

The DØ Collaboration



Panjab U. Chandigarh
Delhi U., Delhi
Tata Institute, Mumbai



University College, Dublin



KDL, Korea U., Seoul
Sungkyunkwan U., Suwon



CINVESTAV, Mexico City



FOM-NIKHEF, Amsterdam
U. of Amsterdam / NIKHEF
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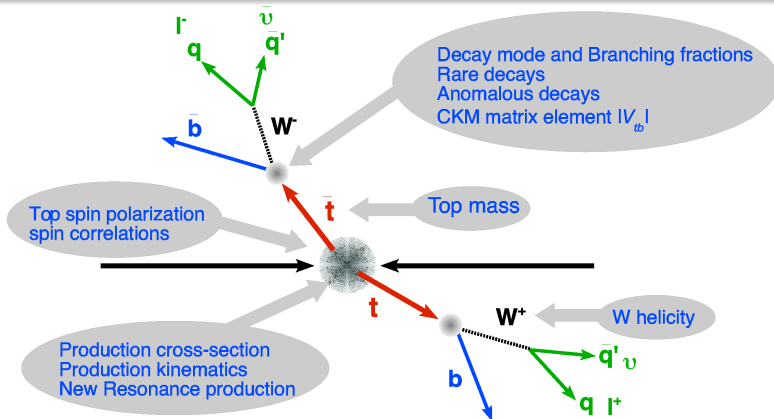
Lancaster U.
Imperial College, London
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Top quark physics

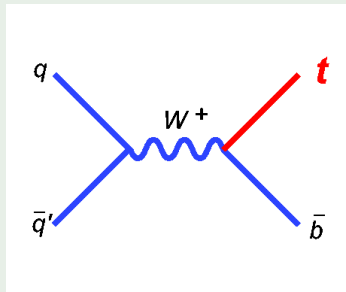
- top quark discovered in 1995 by CDF and DØ at the Tevatron
- Heaviest of all fermions
- Couples strongly to Higgs boson
- So far only observed in pairs, only at the Tevatron



Single top quark production

- Never observed before: electroweak production

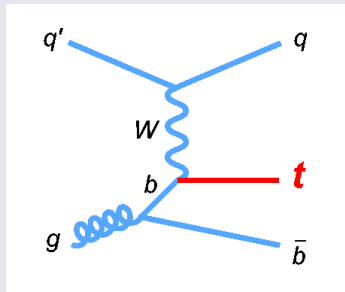
s -channel (tb)



- $\sigma_{NLO} = 0.88 \pm 0.11 \text{ pb} (*)$
- 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.):

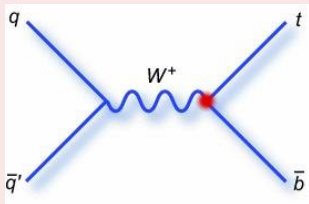
Run II DØ: $< 4.4 \text{ pb} (370 \text{ pb}^{-1})$
Run II CDF: $< 3.2 \text{ pb} (700 \text{ pb}^{-1})$

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



Why do we care? — $|V_{tb}|$

- Has never been observed before!
- Should happen in SM
- The value of the cross section is a SM test and the **first measurement of $|V_{tb}|$**



Direct access to $|V_{tb}|$

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

- 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} :
 $|V_{tb}| = 0.9991^{+0.000034}_{-0.000004}$
- New physics:
 - $V_{td}^2 + V_{ts}^2 + V_{tb}^2 < 1$
 - no constraint on V_{tb}
 - e.g. 4th generation:
 $0.07 < |V_{tb}| < 0.9993$

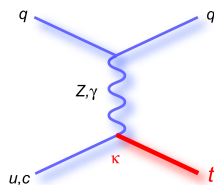
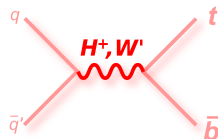


Why do we care? — New physics

- s and t cross sections differently sensitive to new physics

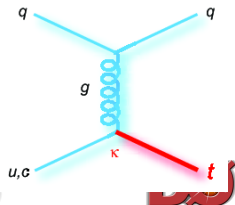
s-channel: charged resonances

- heavy W' boson in topflavour model (separate interaction for 3rd family)
- charged Higgs boson H^\pm in models with extra Higgs doublets (e.g. MSSM)
- charged top pion in topcolor-assisted technicolor
- 4th generation (reduced cross section from $|V_{tb}| < 1$)
- Kaluza-Klein excited W_{KK} , etc...



t-channel: new interactions

- flavour-changing neutral currents ($t-Z/\gamma/g-c$ and/or $t-Z/\gamma/g-u$ couplings)
- 4th generation (potentially strong enhancement from large V_{ts})

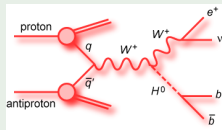


Why do we care? — Spin, Higgs, analysis techniques

Top quark spin

- Large mass \Rightarrow top quark decays before it can hadronize (no top jets)
- First chance to study a bare quark!
- Top polarization reflected in angular distributions of decay products
- SM predicts high degree of left-handed tops \Rightarrow possible sign of new physics, or help pin down what new physics

Higgs searches



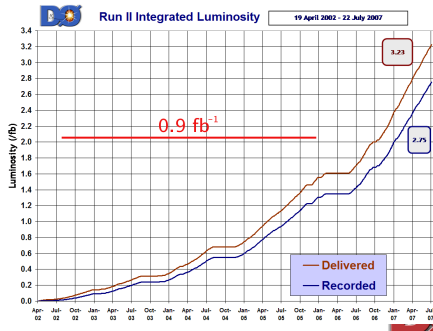
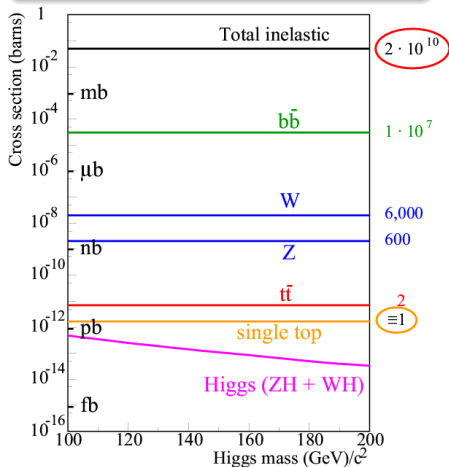
- Important background to WH associated Higgs production
- As soon as we discover it, somebody will try to get rid of it....

Advanced analysis techniques

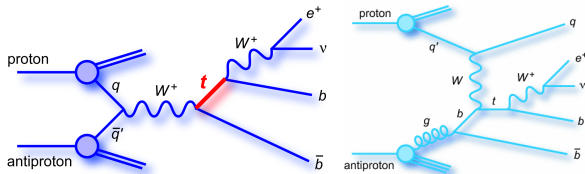
- Test of techniques to extract small signal out of large background
- If tools don't work for single top, forget about the Higgs and other small signals
- If tools don't work at Tevatron, not much hope for LHC

It has been challenging for years...

- Several publications since Run I by DØ and CDF
- 7 DØ and 6 CDF PhDs (Dec '06)
- $\sigma_{t\bar{t}}$ 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

Event selection

- Only one tight (no loose) lepton
 - electron: $p_T > 15 \text{ GeV}$, $|\eta_{det}| < 1.1$
 - muon: $p_T > 18 \text{ GeV}$, $|\eta_{det}| < 2$
- $15 < \cancel{E}_T < 200 \text{ GeV}$
- 2-4 jets: $p_T > 15 \text{ GeV}$, $|\eta| < 3.4$
 - Leading jet: $p_T > 25 \text{ GeV}$, $|\eta_{det}| < 2.5$
 - Second leading jet: $p_T > 20 \text{ GeV}$
- Mis-reconstructed events: require \cancel{E}_T direction not aligned or anti-aligned in azimuth with lepton or jet
- One or two b -tagged jets

Signal and backgrounds

Single top signal ($m_t = 175$ GeV)

- CompHEP-SingleTop + Pythia

W+jets

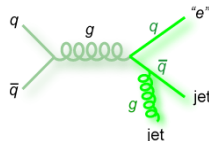
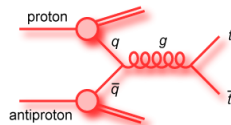
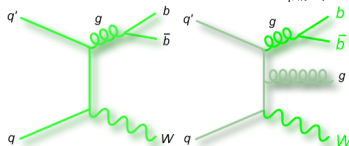
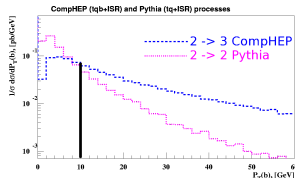
- Most difficult background
- Alpgen+Pythia (MLM matching between matrix elements and parton shower)
- Heavy flavour fraction and normalization from data

$t\bar{t}$ ($m_t = 175$ GeV)

- Alpgen+Pythia (MLM)
- Normalized to $\sigma_{NNLO} = 6.8$ pb

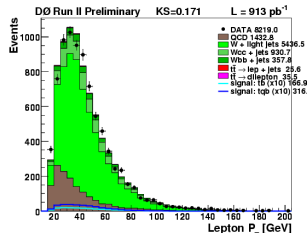
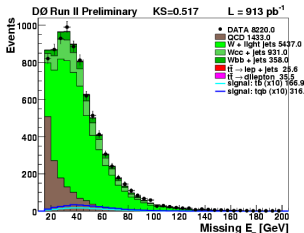
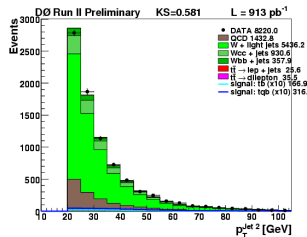
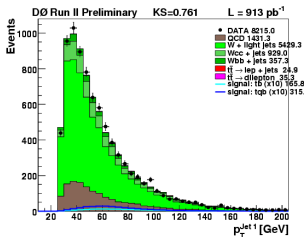
Multijet events

- misidentified lepton, from data



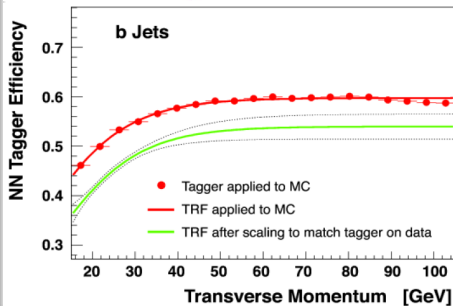
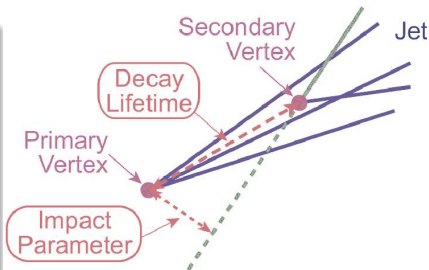
Event selection — Agreement before b tagging

- Normalize W +jets and multijet to data before b tagging
- Checked 90 variables, 4 jet multiplicities, electron + muon
- Good description of data



b-jet tagger

- NN trained on 7 input variables from existing taggers.
 - secondary vertices
 - impact parameter
- 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 η , p_T , z -PV applied to MC
- Operating point:
 - b -jet efficiency $\sim 50\%$
 - c -jet efficiency $\sim 10\%$
 - light jet efficiency $\sim 0.5\%$



Event selection — Splitting by S:B

Percentage of single top <i>tb+tb</i> 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



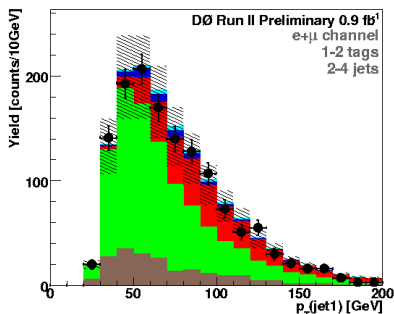
Systematic uncertainties

- 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

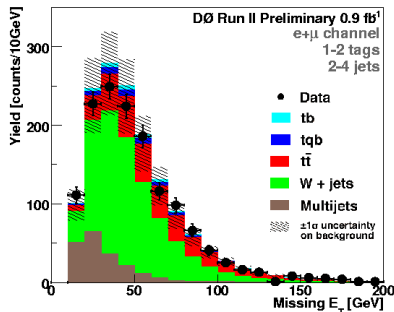
$t\bar{t}$ cross section	18%	Primary vertex	3%
Luminosity	6%	e reco * ID	2%
Electron trigger	3%	e trackmatch & likelihood	5%
Muon trigger	6%	μ reco * ID	7%
Jet energy scale	wide range	μ trackmatch & isolation	2%
Jet efficiency	2%	$\varepsilon_{\text{real}-e}$	2%
Jet fragmentation	5–7%	$\varepsilon_{\text{real}-\mu}$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

Agreement after tagging



Sample	# of Events
s&t-channel Signal	62
Wjj	174
tt→l+jets	266
Wbb & Wcc	675
Mis-ID's leptons	201
Diboson, tt→ dileptons	82

Totals	2 Jets	3 Jets	4 Jets
Data	697	455	246
Total Background	685	460	253
Signal	36	20	6



Multivariate analysis techniques

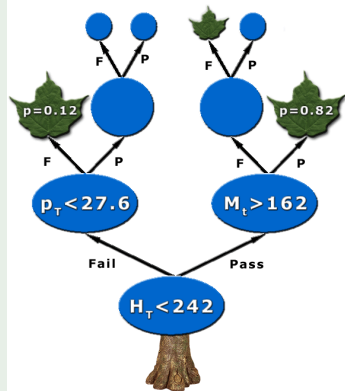
- Boosted decision trees
- Matrix element
- Bayesian neural networks



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
- DT output = leaf purity, close to 1 (0) for signal (bkg)



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



Splitting a node

Impurity $i(t)$

- maximum for equal mix of signal and background
- symmetric in p_{signal} and $p_{\text{background}}$
- 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:

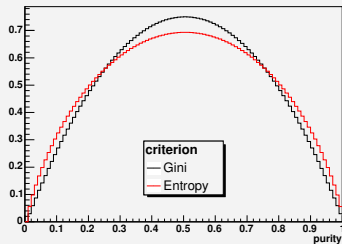
$$\Delta i(s^*, t) = \max_{s \in \{\text{splits}\}} \Delta i(s, t)$$

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

Examples

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

$$\text{entropy} = - \sum_{i=s,b} p_i \log p_i$$



Decision trees — 49 input 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

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

Event Kinematics

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

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



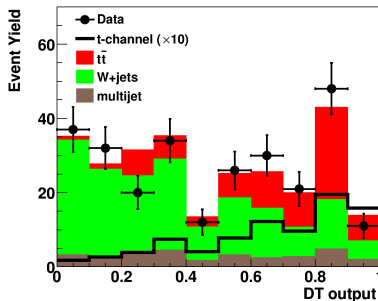
Decision tree output

Measure and apply

- Take trained tree and run on independent pseudo-data sample, determine purities
- Apply to data
- Should see enhanced separation (signal right, background left)
- Could cut on output and measure, or use whole distribution to measure

Limitations

- Instability of tree structure
- Piecewise nature of output



Advantages

- DT has human readable structure (no black box)
- Training is fast
- Deals with discrete variables
- No need to transform inputs
- Resistant to irrelevant variables

Boosting a decision tree

Boosting

- Recent technique to improve performance of a weak classifier
- Recently used on decision trees by GLAST and MiniBooNE
- Basic principle 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 α_k
- Increase weight of misclassified events by e^{α_k}
- Train again to build T_{k+1}
- Boosted result of event i :
$$T(i) = \sum_{k=1}^{N_{\text{tree}}} \alpha_k T_k(i)$$

- 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 tree parameters

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

Analysis strategy

- Train 36 separate trees:
 - 3 signals ($s, t, s + t$)
 - 2 leptons (e, μ)
 - 3 jet multiplicities (2,3,4 jets)
 - 2 b -tag multiplicities (1,2 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 bkgd diagrams to compute event probability density for signal and bkgd hypotheses
- Goal: calculate a discriminant:

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

- Encoded in normalized differential cross section for process S :

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

Used only limited number of Feynman diagrams

- 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)

Bayesian neural networks

A different sort of neural network

- 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
- Used 25 variables (subset of DT variables)
- Same strategy as DT: 36 different BNN

Advantages

- Less prone to overtraining
- Details of each network not important

Limitation

- Darker black box
- Computationally demanding

Implementation: Flexible Bayesian Modeling (FBM) package

<http://www.cs.toronto.edu/~radford/fbm.software.html>

Analysis validation

Ensemble testing

- Test the whole machinery with many sets of pseudo-data
- Like running DØ experiment 1000s of times
- Generated ensembles with different signal contents (no signal, SM, other cross sections, higher luminosity)

Ensemble generation

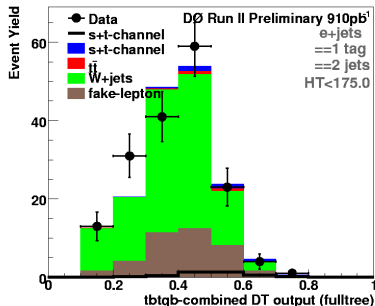
- Pool of weighted signal + background events
- Fluctuate relative and total yields in proportion to systematic errors, reproducing correlations
- Randomly sample from a Poisson distribution about the total yield to simulate statistical fluctuations
- Generate pseudo-data set, pass through full analysis chain (including systematic uncertainties)

All analyses achieved linear response to varying input cross sections and negligible bias

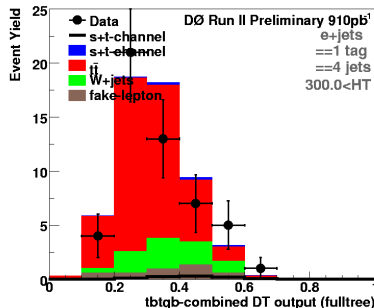
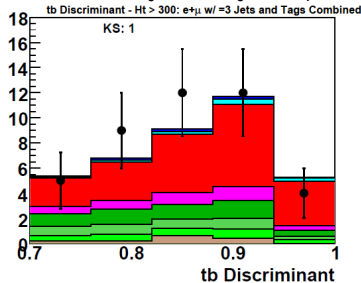


Cross-check samples

- Validate methods on data in no-signal region
- **"W+jets"**: =2jets,
 $H_T(\text{lepton}, \cancel{E}_T, \text{alljets}) < 175 \text{ GeV}$
- **"ttbar"**: =4jets,
 $H_T(\text{lepton}, \cancel{E}_T, \text{alljets}) > 300 \text{ GeV}$
- Good agreement



ME "hard W+jets": =3jets, $H_T > 300 \text{ GeV}$



Sensitivity determination

- Use the 0-signal ensemble

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.

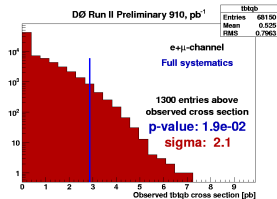
- Also use the SM ensemble to check compatibility of observed result with SM prediction



Expected sensitivity $s+t$

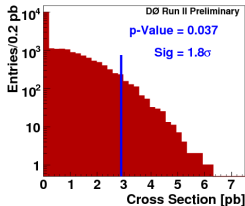
Decision trees

p-value **1.9%** (2.1σ)



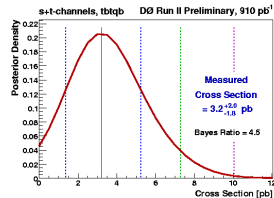
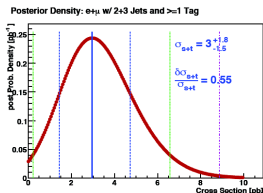
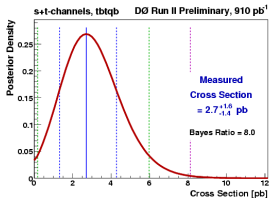
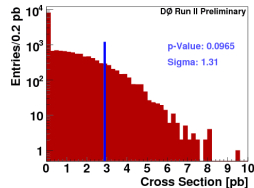
Matrix elements

p-value **3.7%** (1.8σ)



Bayesian NN

p-value **9.7%** (1.3σ)



ME and BNN s+t observed results

Matrix element

$$\sigma = 4.6^{+1.8}_{-1.5} \text{ pb}$$

p-value = 0.21% (2.9σ)

SM compatibility 21%

New preliminary ME result

- Included $t\bar{t} \rightarrow \ell + \text{jets}$ ME in 3-jet discriminant

$$\sigma = 4.8^{+1.6}_{-1.4} \text{ pb}$$

exp. p-value = 3.1% (1.9σ)

obs. p-value = 0.082% (3.2σ)

Bayesian NN

$$\sigma = 5.0 \pm 1.9 \text{ pb}$$

p-value = 0.89% (2.4σ)

SM compatibility 18%

New preliminary BNN result

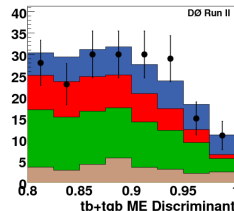
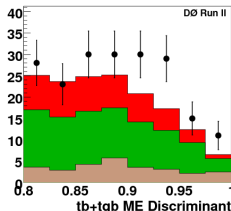
- Better treatment of noisy training data

$$\sigma = 4.4^{+1.6}_{-1.4} \text{ pb}$$

exp. p-value = 1.6% (2.2σ)

obs. p-value = 0.083% (3.1σ)

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



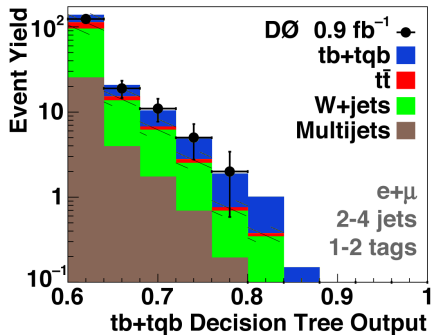
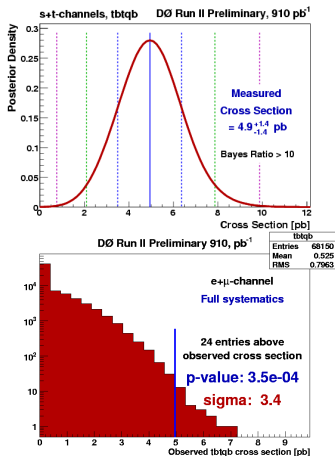
Boosted decision tree observed results

$$\sigma_{s+t} = 4.9 \pm 1.4 \text{ pb}$$

$$p\text{-value} = 0.035\% (3.4\sigma)$$

SM compatibility: 11% (1.3σ)

Evidence for single top production!



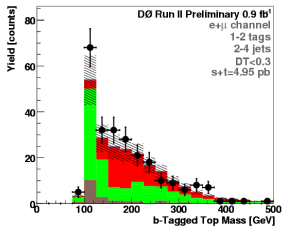
$$\sigma_s = 1.0 \pm 0.9 \text{ pb}$$

$$\sigma_t = 4.2^{+1.8}_{-1.4} \text{ pb}$$

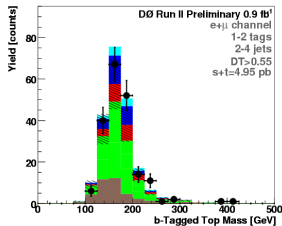


Boosted decision tree event characteristics

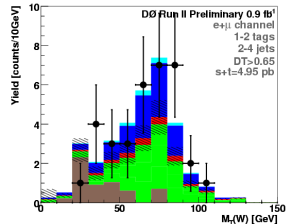
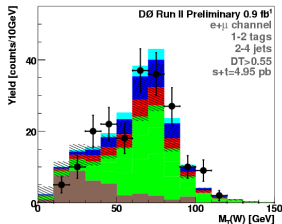
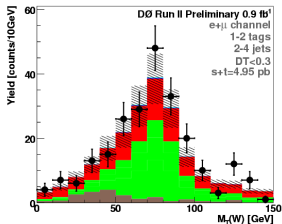
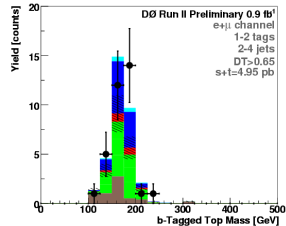
$DT < 0.3$



$DT > 0.55$



$DT > 0.65$



Measuring $|V_{tb}|$

- Now that we have a cross section measurement, we can make the first direct measurement of $|V_{tb}|$
- Use the same infrastructure as for cross section measurement but make a posterior in $|V_{tb}|^2$

Additional theoretical errors (hep-ph/0408049)

	s	t
top mass	13%	8.5%
scale	5.4%	4.0%
PDF	4.3%	10.0%
α_s	1.4%	0.01%

- Most general Wtb coupling ($P_{L,R} = (1 \mp \gamma_5)/2$):

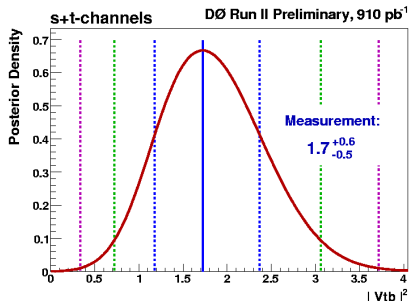
$$\Gamma_{tbW}^\mu = -\frac{g}{\sqrt{2}} V_{tb} \bar{u}(p_b) \left[\gamma^\mu (f_1^L P_L + f_1^R P_R) - \frac{i\sigma^{\mu\nu}}{M_W} (f_2^L P_L + f_2^R P_R) \right] u(p_t)$$

- SM: $f_1^L = 1$, $f_1^R = 0$ (pure $V-A$), $f_2^L = f_2^R = 0$ (CP conservation)
- Effectively measuring strength of $V-A$ coupling $|V_{tb} f_1^L|$, can be > 1

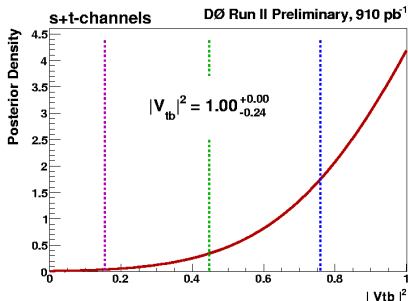


First direct measurement of $|V_{tb}|$

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



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



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

(assuming $f_1^L = 1$, flat prior in $[0,1]$)

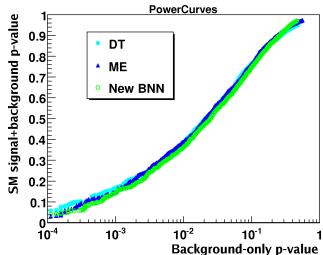
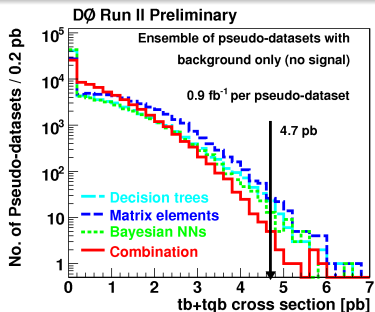
- No assumption about number of quark families or CKM matrix unitarity

New: combination of s+t analyses

Correlations

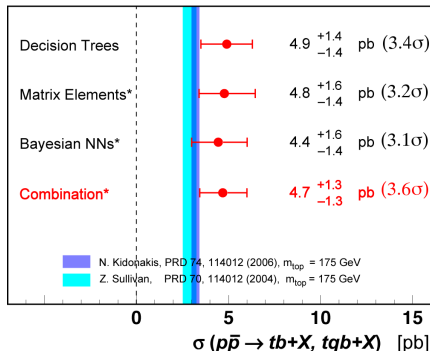
- 3 analyses with similar performance on same dataset
- Combined using BLUE method

	DT	ME	BNN
DT	100%	64%	66%
ME		100%	59%
BNN			100%



DØ Run II * = preliminary

0.9 fb⁻¹



Conclusion

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 σ 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)

Published in Phys. Rev. Lett. 98, 181802 (2007) (hep-ex/0612052)

New preliminary combination of DT, ME and BNN

$$\sigma(p\bar{p} \rightarrow tb + X, tqb + X) = 4.7 \pm 1.3 \text{ pb}$$

3.6 σ significance

- A lot more data already at hand



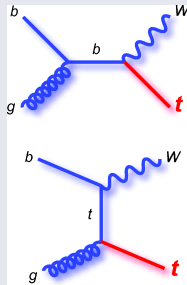
Single top prospects — Tevatron and LHC

Tevatron

- By 2009 we should have observed single top production and measured its cross section to 15-20%
- $|V_{tb}|$ is then known to $\sim 10\%$

LHC

- Much larger production rates:
 $\sigma_s^{t/\bar{t}} = 6.6/4.1 \text{ pb } (\pm 10\%)$
 $\sigma_t^{t/\bar{t}} = 156/91 \text{ pb } (\pm 5\%)$
 $\sigma_{tW}^{t/\bar{t}} = 34/34 \text{ pb } (\pm 10\%)$
- Try to observe all three channels (s-channel challenging)
- $|V_{tb}|$ measured to percent level
- Large samples \Rightarrow study properties



- More information:

<http://www-d0.fnal.gov/Run2Physics/top/public/fall06/singletop>

