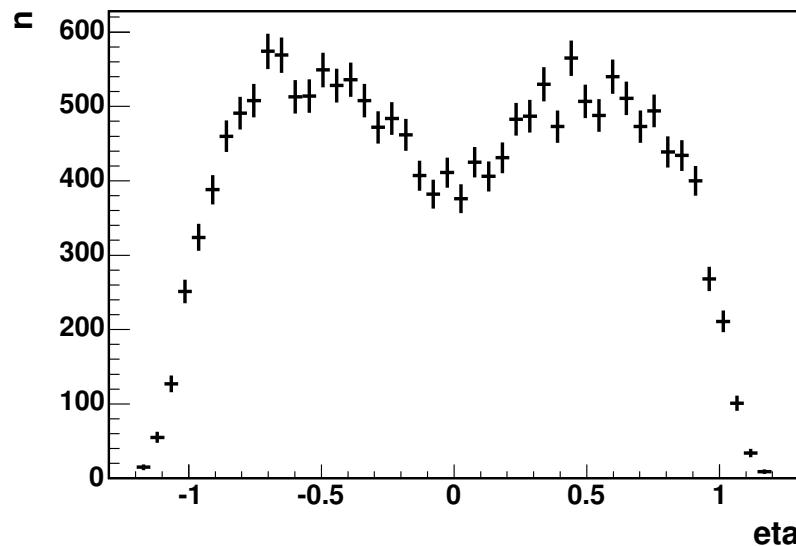


An Artificial Neural Network for Plug Electron Identification

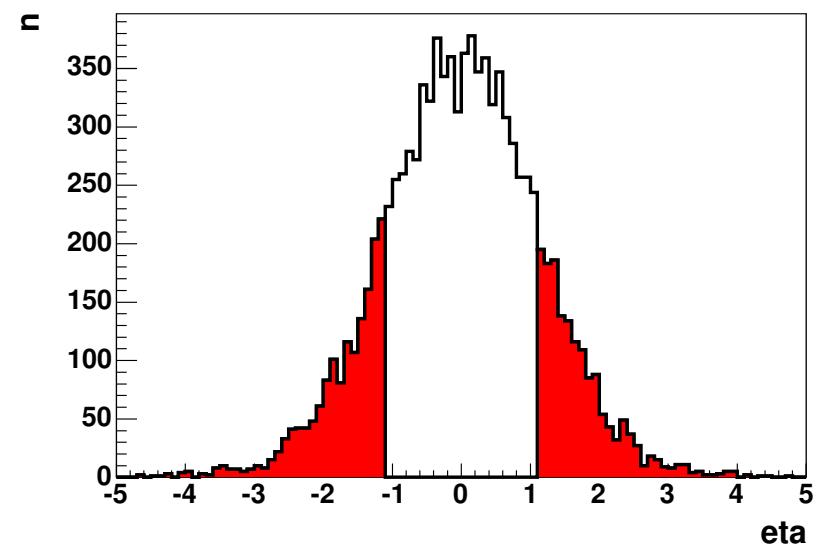
Yves Kemp, Thomas Müller, Hartmut Stadie, Wolfgang Wagner

Universität Karlsruhe

Use of plug electrons for analysis



η -distribution of electrons
in 4.11 version of single top
analysis
(after preselection)

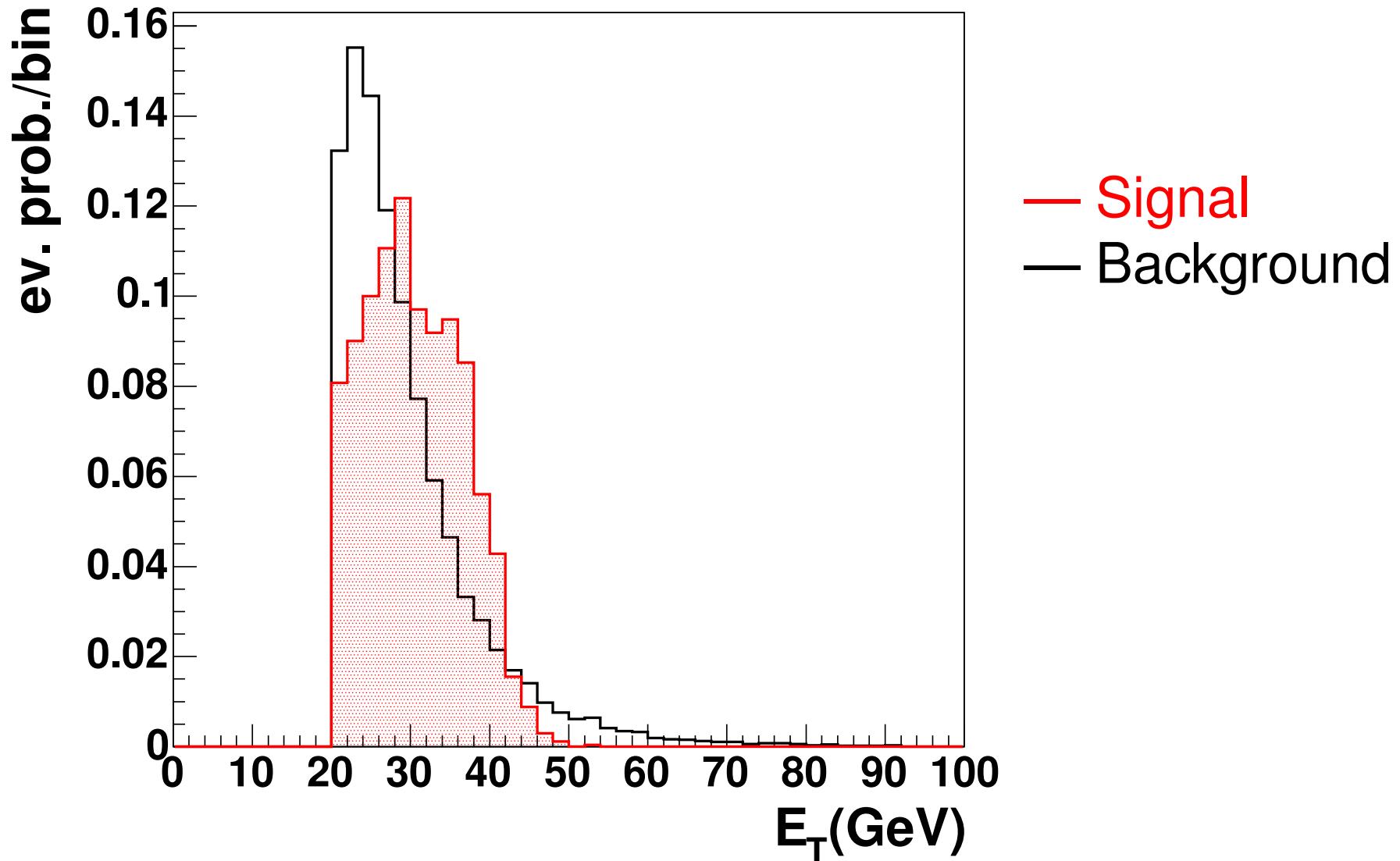


η -distribution of electrons
in t-channel MC.
ca 30% with $|\eta| > 1.1$

Selection of the samples

- Signal sample:
 - 1 tight central electron with track
 - Another electron candidate in plug (Z-Candidate)
 - Cuts to be independent of trigger cuts
 - 2000 events remain
- Background sample:
 - 2 balanced jets (1 central, 1 plug)
 - Several preselection cuts
 - 30000 events remain
- Both samples taken from data! (bpel08)
cdfsoft2 versions 4.9 and 4.11

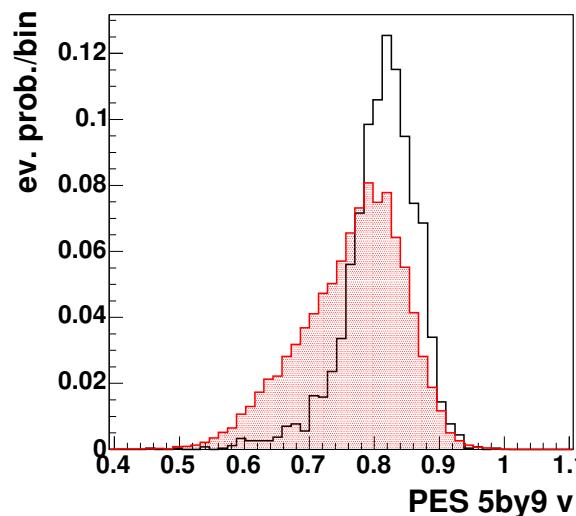
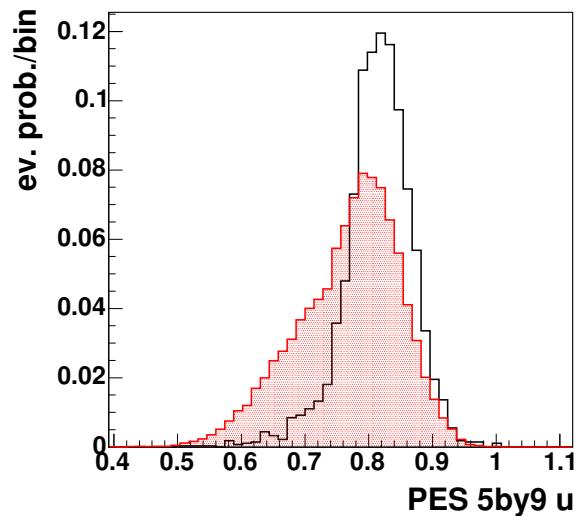
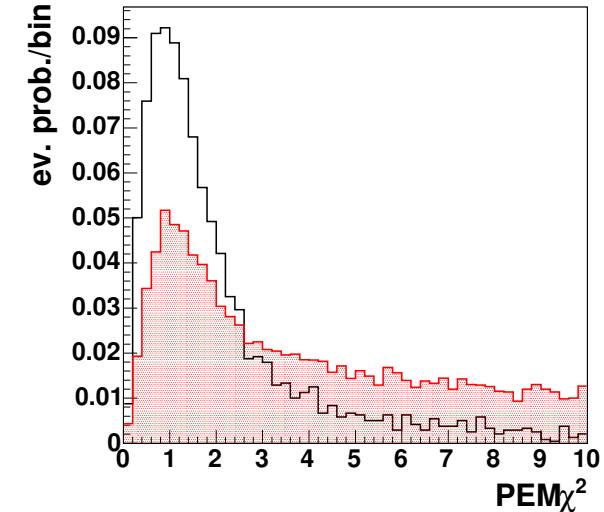
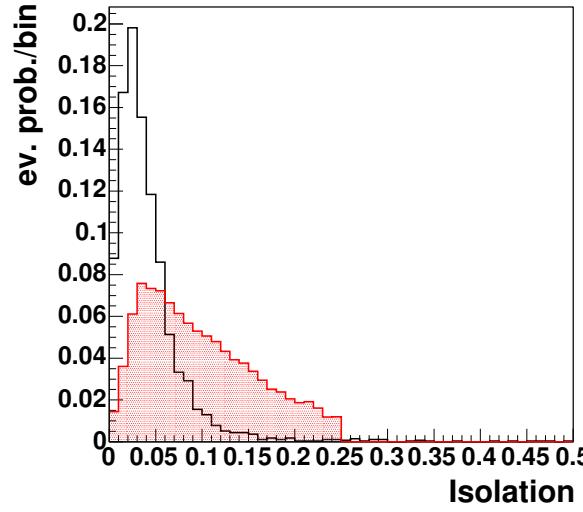
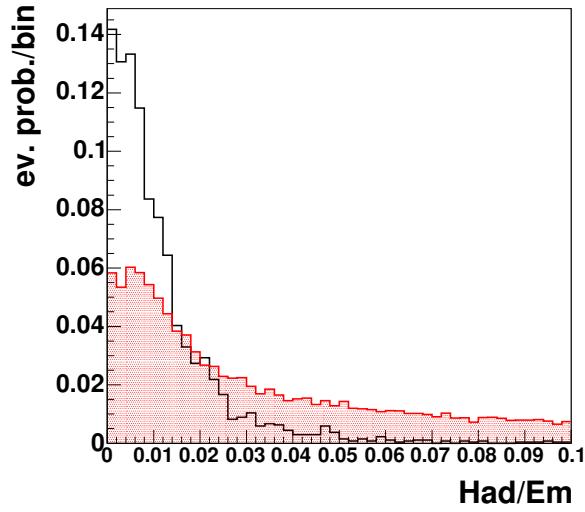
Control plot: E_T of plug electron



Variables for plug electron ID

- Fiducial cut: $1.2 < |\eta| < 2.8$
- EmE_T
- (Phoenix-)Track
- HadE/EmE (sliding cut)
- Isolation Ratio
- PEM χ^2 (comparision with test beam data)
- PES 5by9 u (Shower profile in PES in u direction)
- PES 5by9 v (Shower profile in PES in v direction)

Selection variables

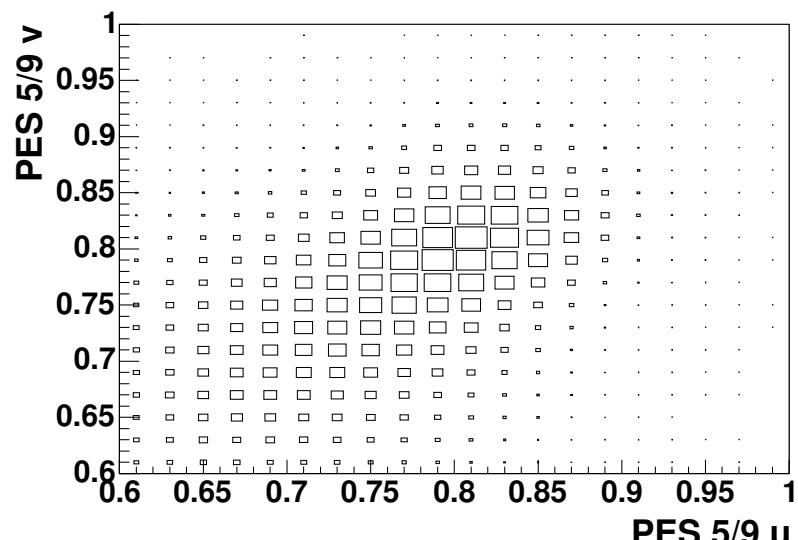


— Signal
— Background
(normalized to same integral)

Correlation matrix

	Target	HadEm	Iso	$\text{PEM}\chi^2$	PES 5/9 u	PES 5/9 v
Target	100.0	-49.4	-66.6	-64.5	43.9	43.2
HadEm		100.0	52.8	44.8	-24.8	-24.2
Iso			100.0	72.0	-38.9	-38.5
$\text{PEM}\chi^2$				100.0	-42.8	-43.4
PES 5/9 u					100.0	45.9
PES 5/9 v						100.0

Target is -1 for background, 1 for signal



Correlation between the two PES 5/9 variables due to cross talk and geometry

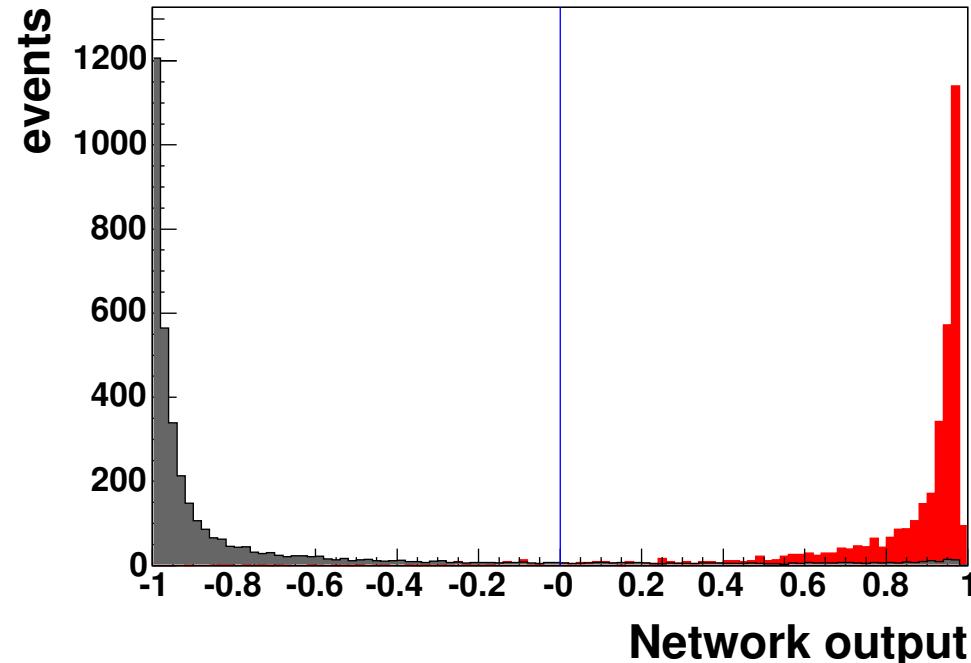
Artificial Neural Network

- 5 input variables (+ 1 bias node)

	rank	correlation σ
Had/Em	5	4.8
Isolation	1	57
PEM χ^2	3	14
PES 5/9 u	4	11
PES 5/9 v	2	20

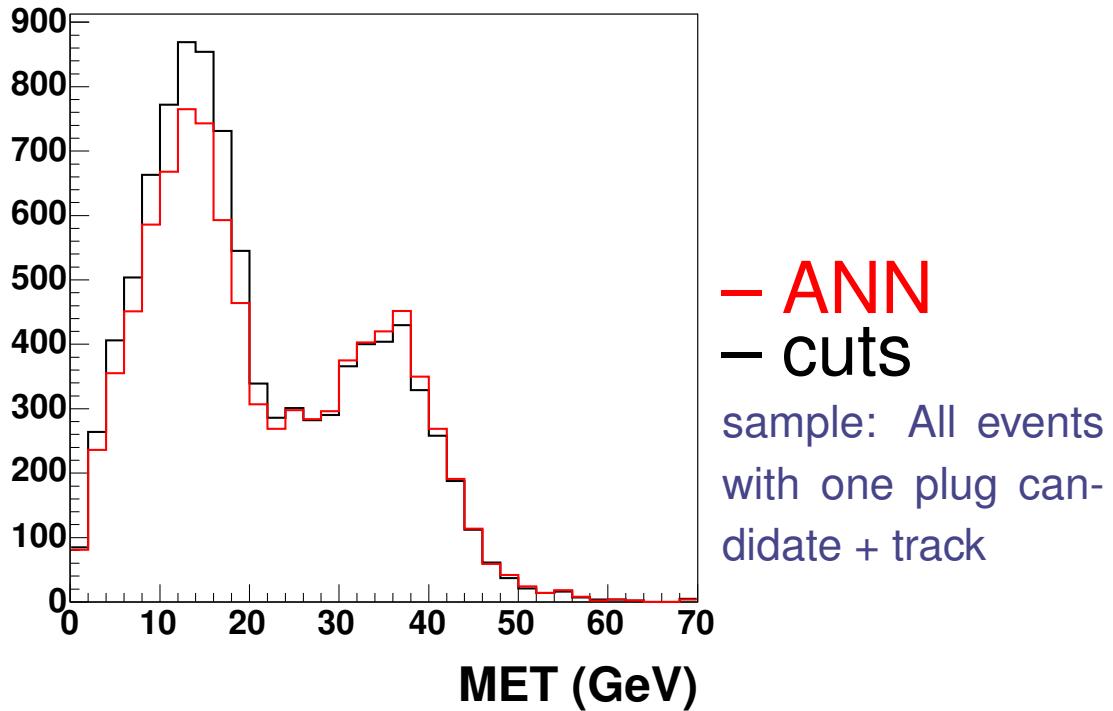
- Preprocessing of the variables
- 5 nodes in intermediate layer
- Binary classification (-1 background, 1 signal)
- 50 iterations
- NeuroBayes ANN package (see Ulrichs talk)

Artificial Neural Network results

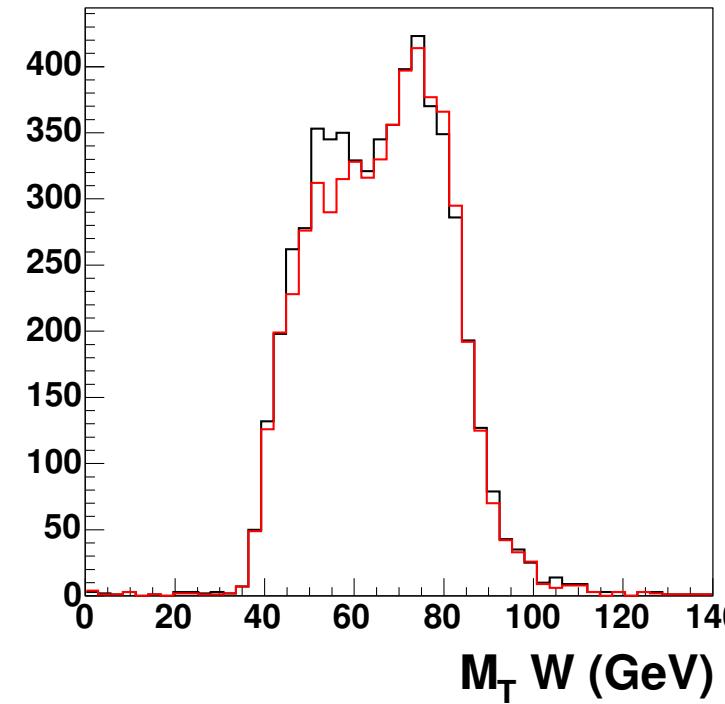


- Standard Cuts:
 - Signal efficiency 84%
 - Background eff. 5.3%
- ANN cut > 0.65 :
 - Same signal efficiency
 - Background eff. 3.5%
 - 34% less background
- ANN cut > 0.40 :
 - Same background eff.
 - Signal efficiency 91%
 - 8% more signal

Independent tests

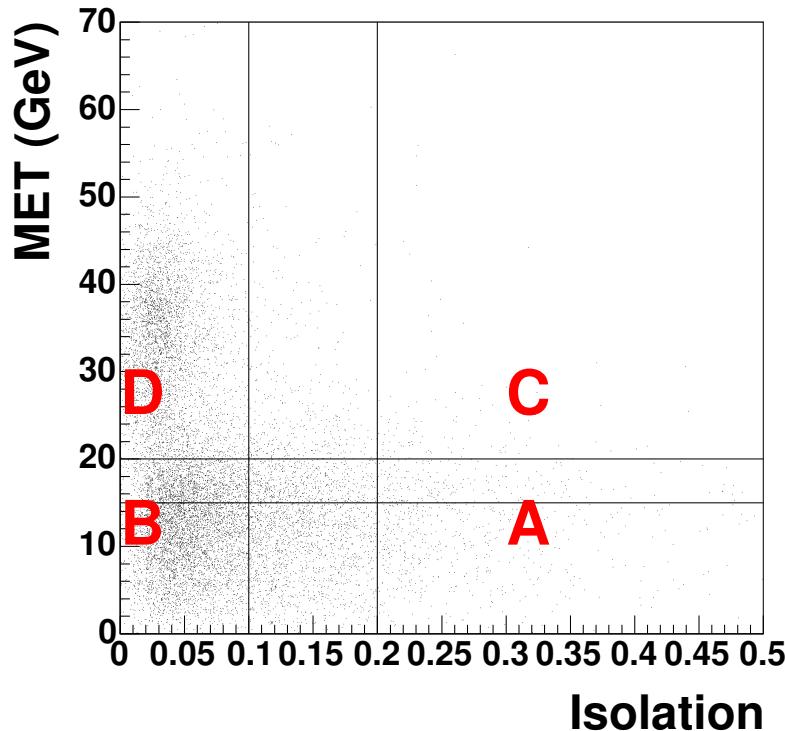


Missing ET
NN cut: 5679 ev. < 25 GeV
standard: 6483 ev. < 25 GeV



Transverse W-Mass
($MET > 15$ GeV)
NN cut: 5738 events
standard: 5533 events

Background estimation: 4-sector method



All standard cuts except isolation applied

- A: ($\text{MET} < 15 \text{ GeV}$, $\text{Iso} > 0.2$): 501
- B: ($\text{MET} < 15 \text{ GeV}$, $\text{Iso} < 0.1$): 4015
- C: ($\text{MET} > 20 \text{ GeV}$, $\text{Iso} > 0.2$): 106
- D: ($\text{MET} > 20 \text{ GeV}$, $\text{Iso} < 0.1$): 4156

$$\Rightarrow \frac{N_{\text{QCD in D}}}{N_{\text{Total in D}}} = 20.4\%$$

This method does not work for ANN cut

sample: All events with one plug candidate + track

QCD background estimation: Fit method

Idea: Fit background and MC to data

Background: Invert ANN cut: < -0.95

MC: $W+jets$ sample

Fit: Scaling factors for background and MC, fit MET in interval 15-30 GeV

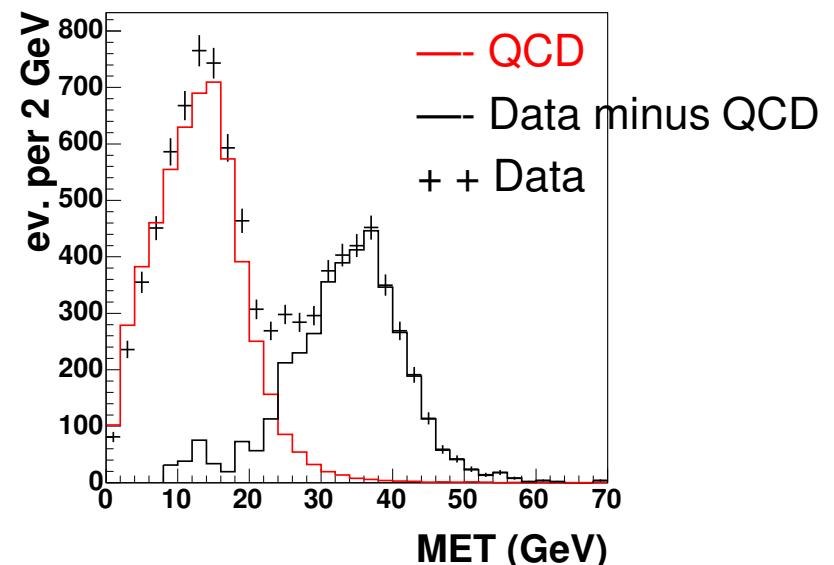
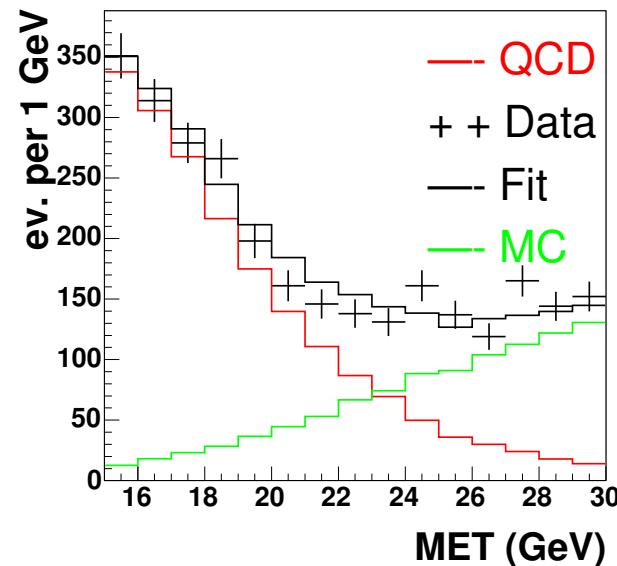
ANN > 0.5 :

$$\Rightarrow \frac{N_{\text{QCD}} \text{ in MET} > 20 \text{ GeV}}{N_{\text{Total}} \text{ in MET} > 20 \text{ GeV}} = 15.3\%$$

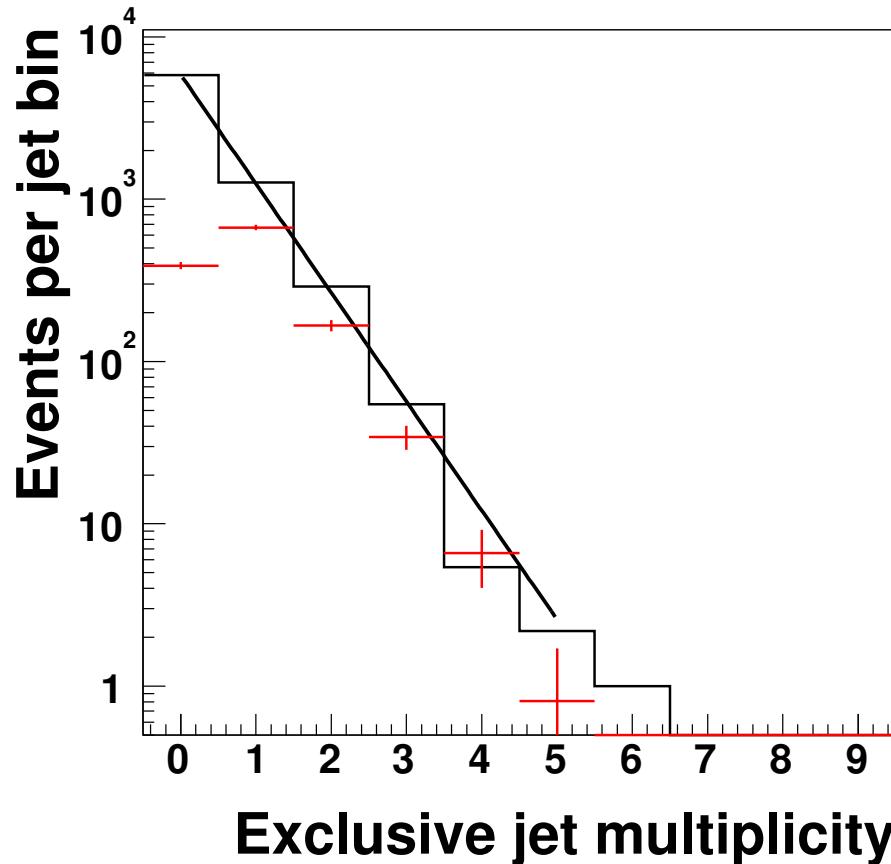
Standard Cuts:

$$\Rightarrow \frac{N_{\text{QCD}} \text{ in MET} > 20 \text{ GeV}}{N_{\text{Total}} \text{ in MET} > 20 \text{ GeV}} = 17.9\%$$

sample: All events with one plug candidate + track



Jet multiplicity



++ QCD (ANN < -0.95)
— Data minus QCD

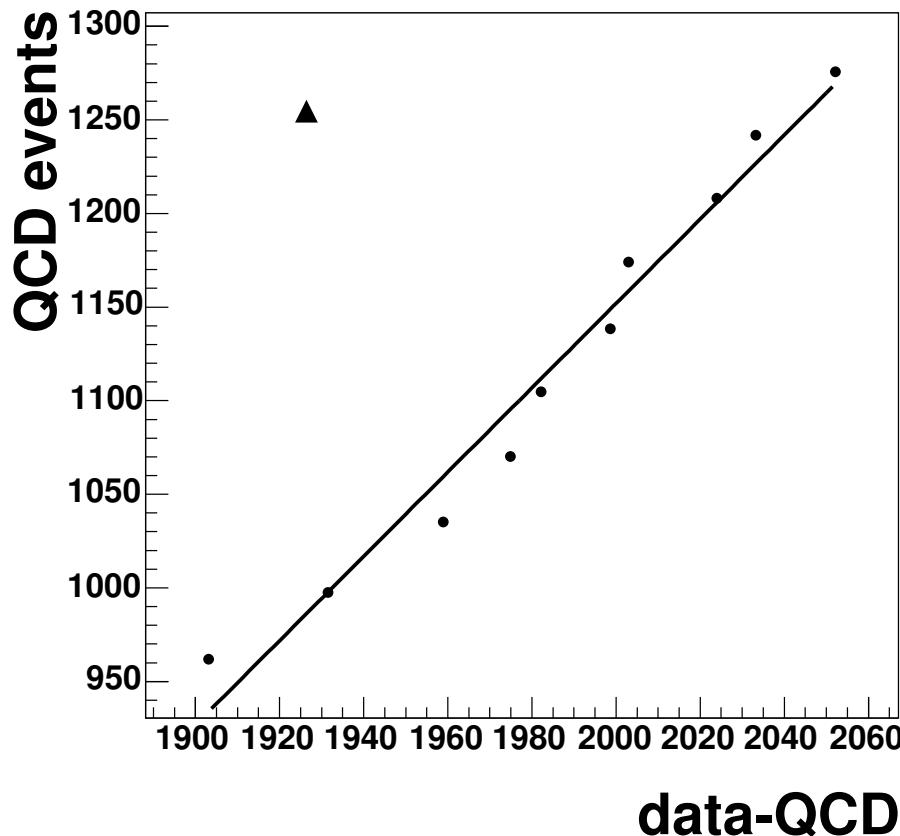
Jets	Data	QCD	Data - QCD
0	39693	2581	37112
1	7507	4204	3303
2	1976	1083	893
3	402	240	262
4	53	44	11

With CDF standard cuts

0	37960	2990	35870
2	1926	1255	673

sample: All events with one plug candidate + track
Additional requirement: MET > 20 GeV
(realistic scenario for W+jets analysis)

Exclusive 2-Jet Bin



- Varying ANN cut such that S/B or S/\sqrt{B} is optimal
- $\sim 20\%$ less background
- $\sim 5.7\%$ more signal

Conclusion, Outlook

- Plug electrons will give better single top limit
 - Identification with Artificial Neural Network useful
 - ~20% less background in 2-jet-bin
 - ~5.7% more signal in 2-jet-bin
 - Plan to provide easy-to-use generic tool
-
- Waiting for Gen 5 plug TopNtuples
 - Determine acceptance and efficiencies
 - Integrate into next round of single top analysis