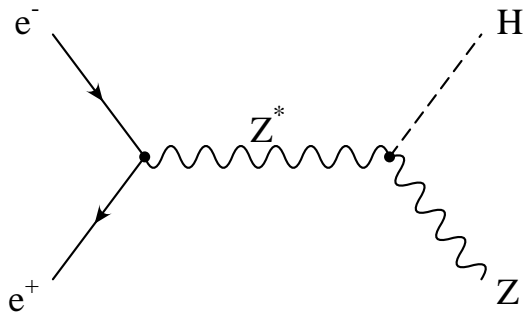


$ZH \rightarrow q\bar{q}b\bar{b}$ study with neural network

David Ward and Wenbiao Yan



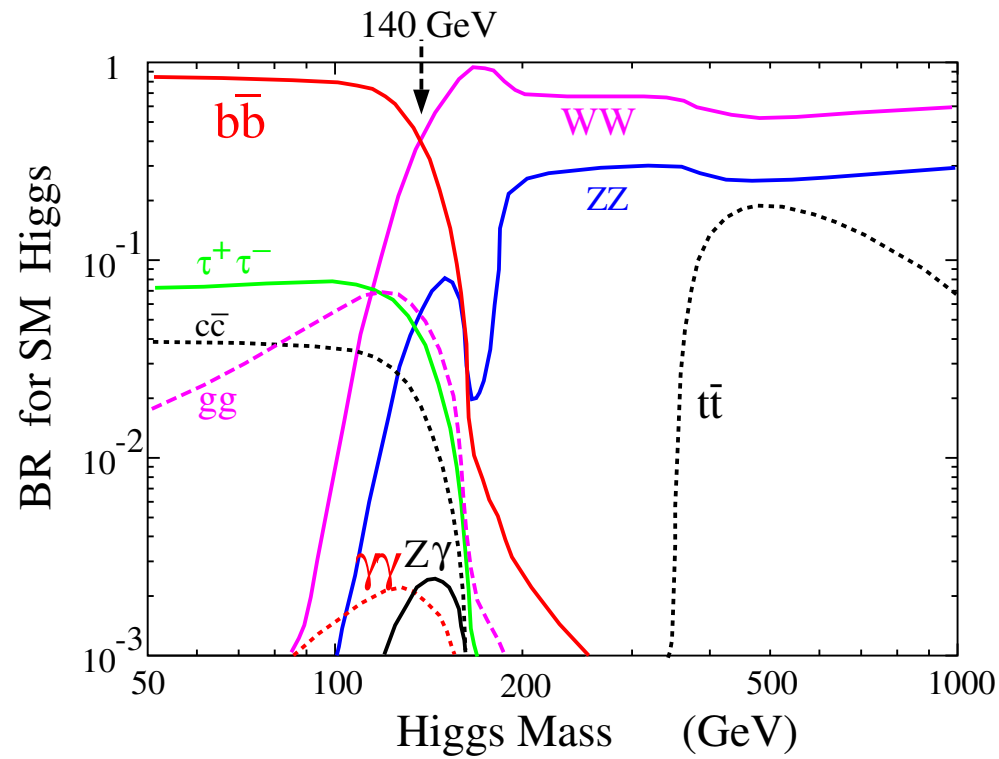
- $ZH \rightarrow q\bar{q}b\bar{b}$



- $Z \rightarrow q\bar{q}; H \rightarrow b\bar{b}$

- introduction
- jet energy correction
- jet resolution
- neural network @ ZH

Higgs decay



- $\text{Br}(h \rightarrow b\bar{b}) \sim 70\% \text{ @ } M_h = 120 \text{ GeV}$
- Higgs mass reconstruction via $M_{b\bar{b}}$

MC data samples @ 250 GeV

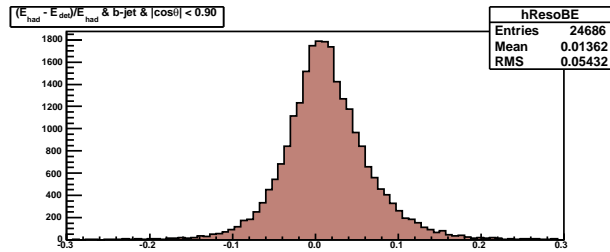
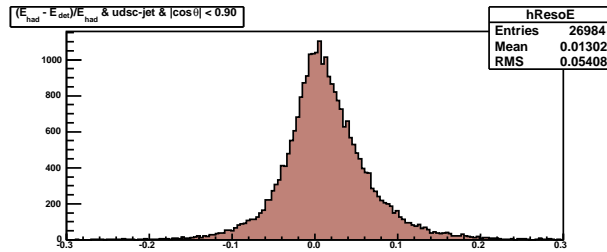
- **Pythia without Beamstrahlung** \implies move to mass production samples when they are available

	σ^{my} (fb)	event number	σ^{Akiya} (fb)
$ZH \rightarrow q\bar{q}H$	159.5	40K	157.5
$WW \rightarrow q_1q_2q_3q_4$	7146.84	1786.7K	7109.84
$ZZ \rightarrow q_1\bar{q}_1q_2\bar{q}_2$	588.1	147.3K	586.0
$e^+e^- \rightarrow q\bar{q}$	47682.6	11920K	

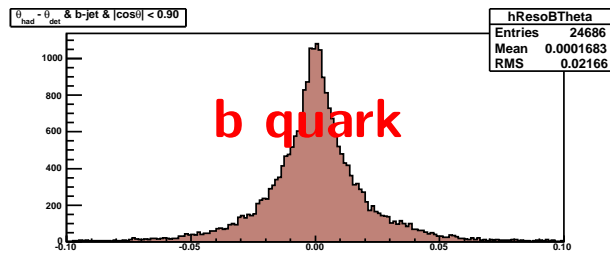
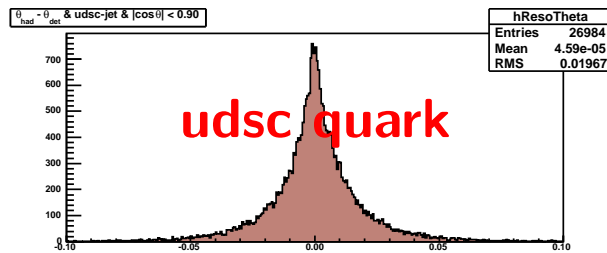
- $ZH \rightarrow q\bar{q}H$: Pythia $\sim 40K$ (signal)
 $ZH \rightarrow q\bar{q}b\bar{b}$: Pythia $\sim 80K$ (training & test @ Neural network)
- $WW \rightarrow q_1q_2q_3q_4$: Pythia $\sim 300K$
 $ZZ \rightarrow q_1\bar{q}_1q_2\bar{q}_2$: Pythia $\sim 148K$
 QQ: Pythia $\sim 500K$
 QQ: 20/40/80/91.2/120/160/200/240/300 GeV (jet resolution) $\sim 9*30K$
- LDCPrime_02Sc @ Mokka06-06-p03: (SAME binary package for DESY mass production)
 event reconstruction ilcinstall v01-04

Jet resolution

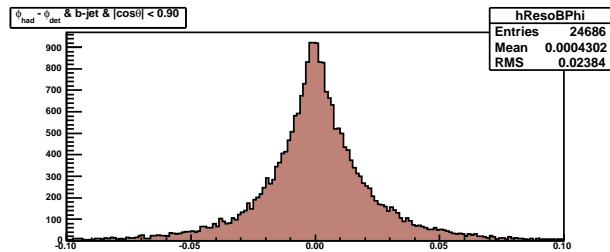
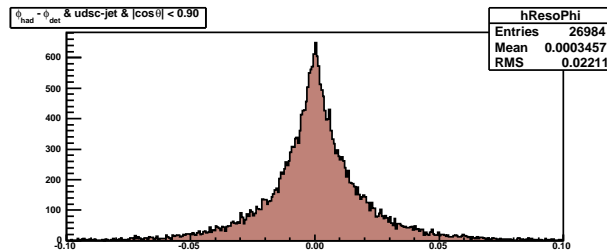
E



θ



ϕ



- Jets finding at hadron level: MC stable particles without neutrinos
- Jets finding at detector level
- Jet's energy: mean value $\sim 1.3\%$ \rightarrow jet energy correction
- Jet's angle: mean value ~ 0.0

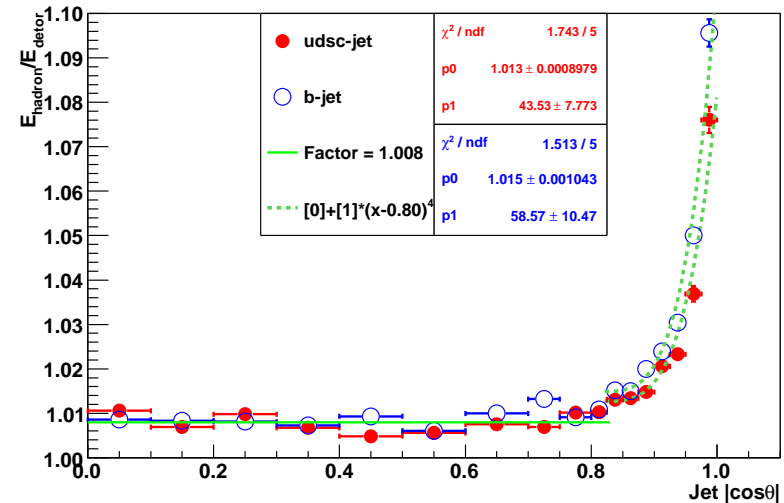
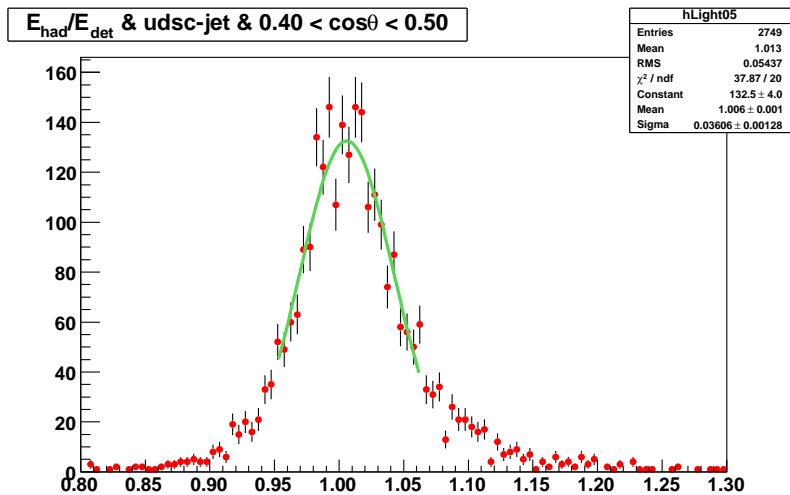
Jet energy correction

Assuming the direction vector of reconstructed jets correctly measured with that of quarks, the correction factors are scale factor on the jet.

- **Step I: Correct for calibration constants and detector effects**
 - Jets at hadron level: MC stable particles without neutrinos
 - Jets at detector level
- **Step II: Correct for neutrinos for udsc-jets**
 - use $Z^0 \rightarrow q\bar{q}$ ($q = u, d, s, c$) events, let $M_{Z^0} = 91.2$ at detector level
- **Step III: Correct for neutrinos for b-jets**
 - use identified leptons inside jets

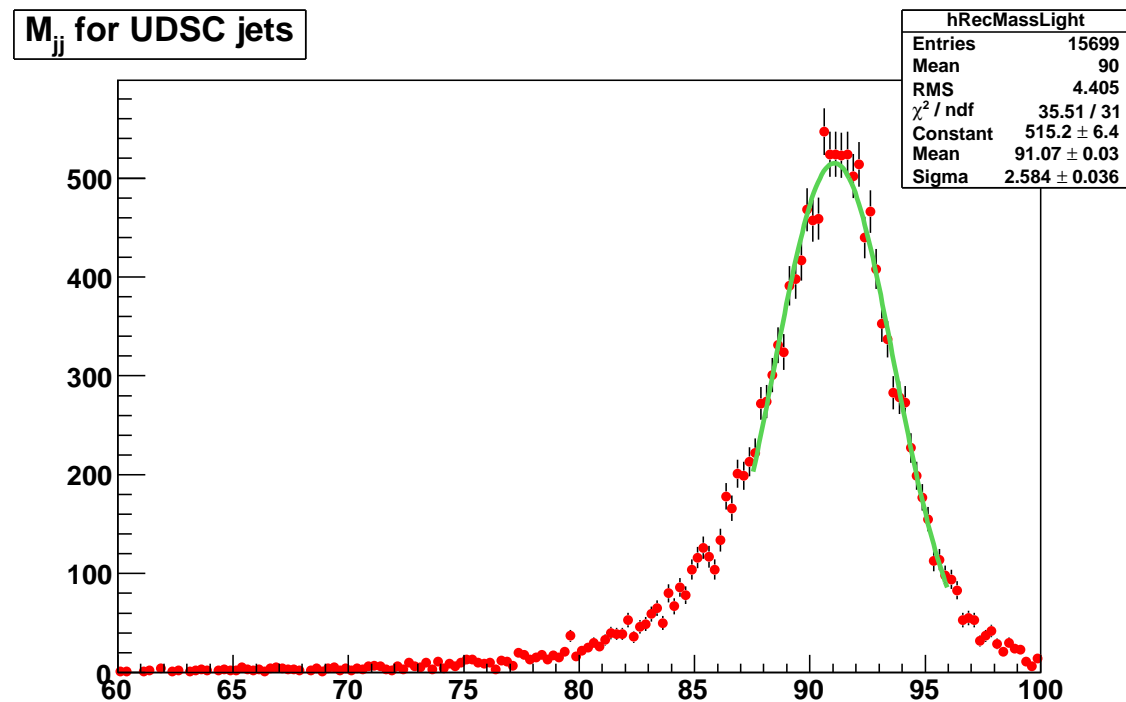
Jet energy correction: Step I

- Correct for calibration constants and detector effects
 - Jets at hadron level: MC stable particles without neutrinos
 - Jets at detector level
- $E_{hadron}/E_{detector}$ for udsc-jet & b-jet



Jet energy correction: Step II

- neutrinos contribution to udsc-jets: using udsc-jets from Z^0
- udsc-jets: $91.2/91.07 = 1.0014$; global factor

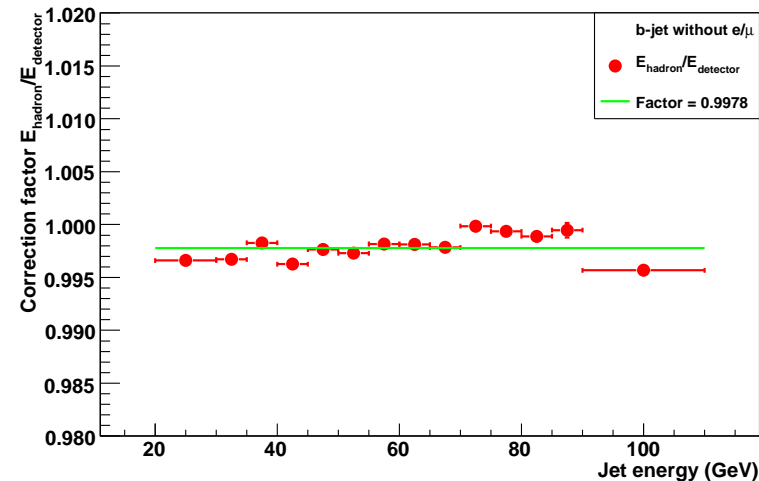
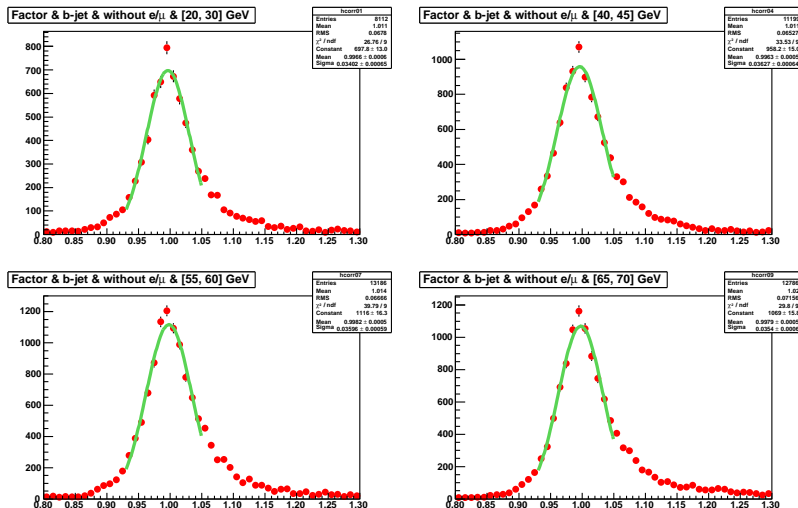


Jet energy correction: Step III

- identified leptons (e/μ) inside jets to correct for neutrinos
- direct decays $b \rightarrow l + X$ and cascade decays $b \rightarrow c \rightarrow l + X$
 - particles at detector level with LCRelation RecoMCTruthLink \rightarrow particles at hadron level \rightarrow identified leptons with PDG code
 - identified leptons from B-hadron or D-hadron
 - identified leptons ≥ 2 : sum of several identified leptons
 - lepton energy: $E_{lepton} > 4.0\text{GeV}$
 - $0.95 < E_{lepton}^{rec}/E_{lepton}^{MC} < 1.05$
- selecte b-jets with identified leptons for energy correction
 - lepton energy: $E_{lepton} > 4.0\text{GeV}$
 - $0.95 < E_{lepton}^{rec}/E_{lepton}^{MC} < 1.05$
 - $|\cos \theta_{jet}| < 0.80$

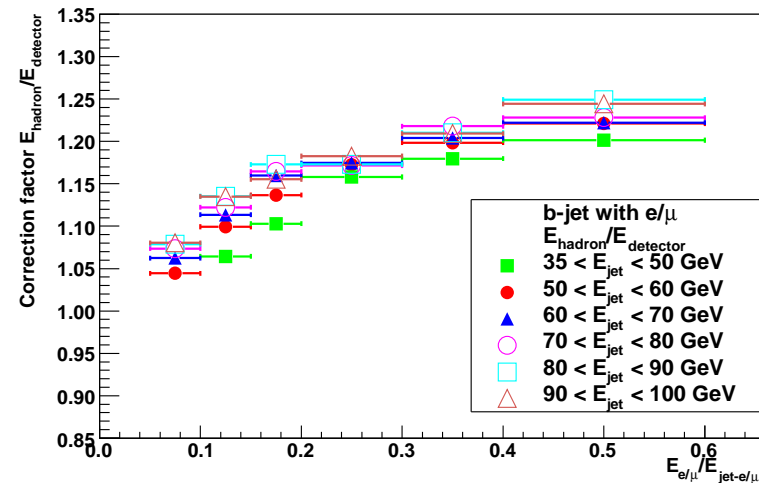
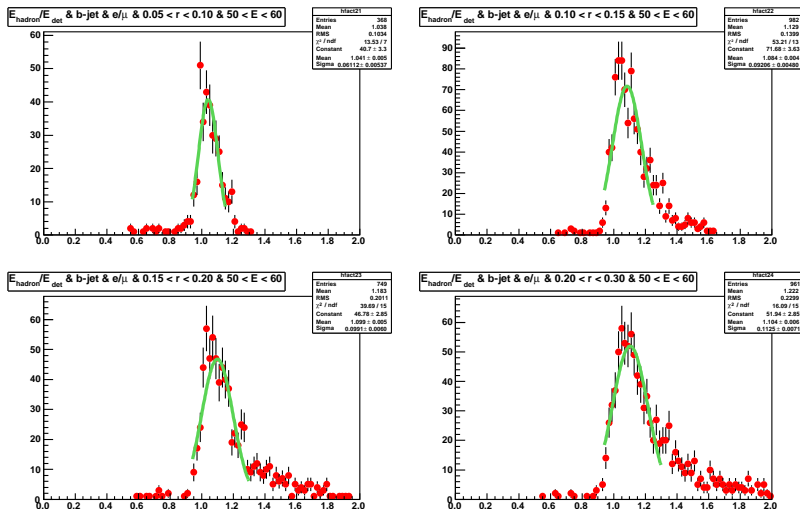
Jet energy correction: b-jet without e/μ

- correction factor $E_{hadron}^{jet}/E_{detector}^{jet}$
 - Jets at hadron level: MC stable particles (**with neutrinos**)
 - Jets at detector level
- Gaussian fitting: (**not bad**)
- long tail from τ neutrinos
- mean value from Gaussian fitting
- b-jet correction: average 0.9978



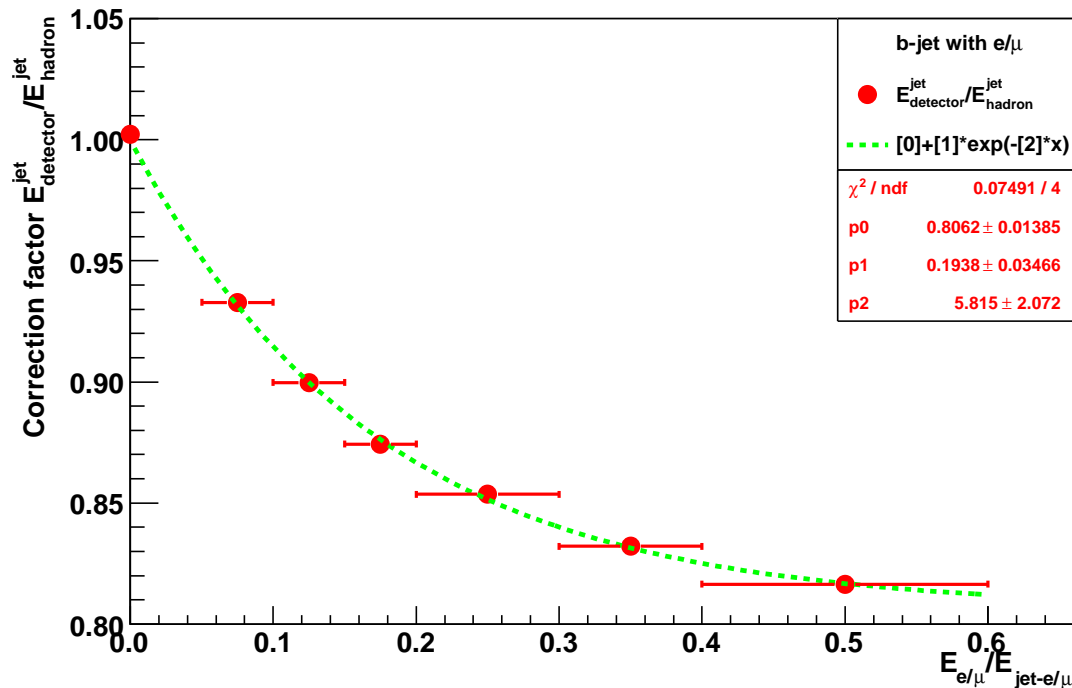
Jet energy correction: b-jet with e/μ

- correction factor $E_{hadron}^{jet}/E_{detector}^{jet}$
 - Jets at hadron level: MC stable particles (**with neutrinos**)
 - Jets at detector level
- Gaussian fitting: (**not good**)
- $E_{e/\mu}$: lepton energy
- $E_{jet-e/\mu}$: $E_{jet} - E_{e/\mu}$

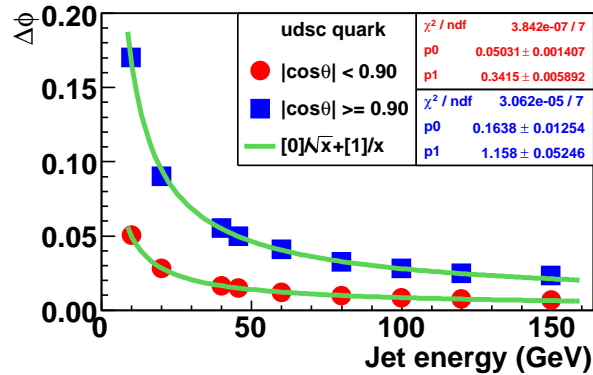
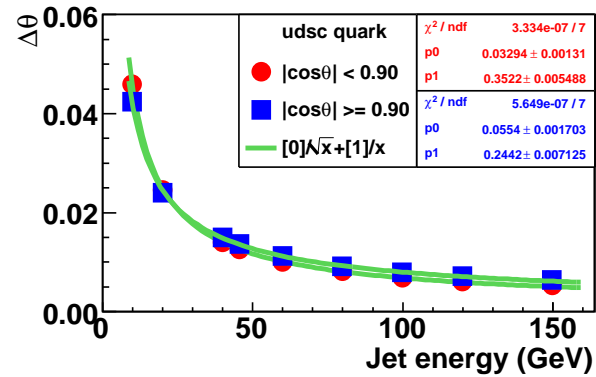
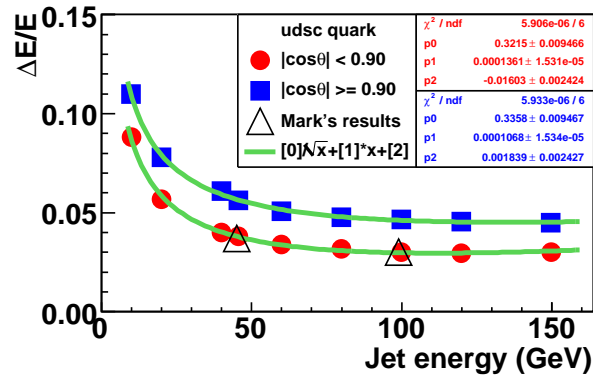


Jet energy correction: b-jet with e/μ

- correction factor $E_{detector}^{jet}/E_{hadron}^{jet}$ as function of $E_{e/\mu}/E_{jet-e/\mu}$
- point (0.0, 1.0/0.9978): b-jet without e/μ

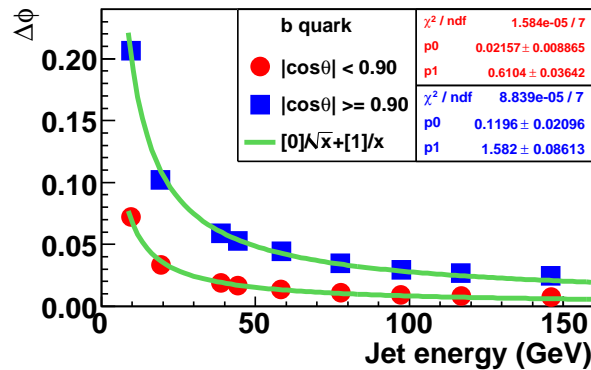
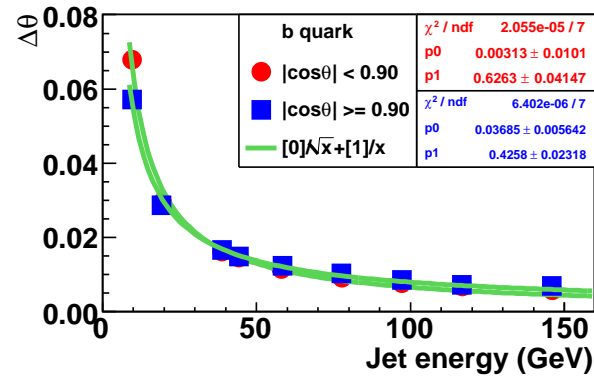
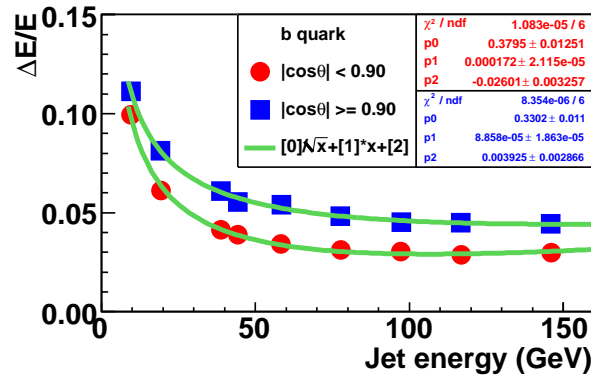


Jet resolution: udsc-jet



- $e^+e^- \rightarrow q\bar{q}$ ($q = u, d, s, c$) without ISR
- Jets finding at hadron level: MC stable particles without neutrinos

Jet resolution: b-jet

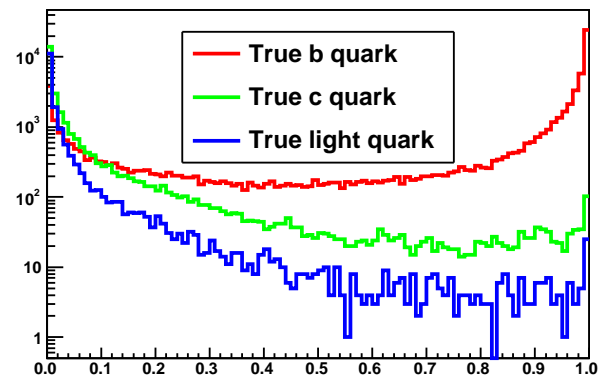


- $e^+e^- \rightarrow q\bar{q}$ ($q = b$) without ISR
- Jets finding at hadron level: MC stable particles without neutrinos

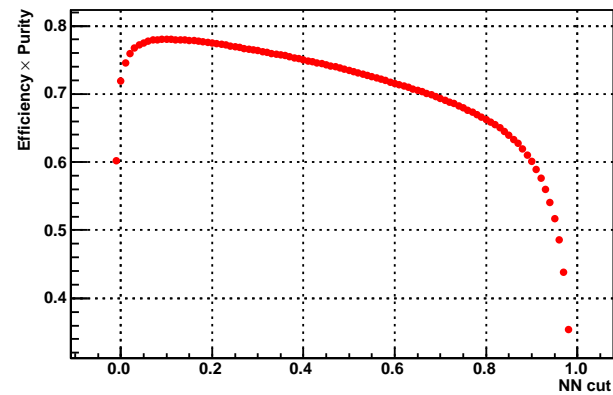
B-tag at ZH 250GeV

- B-tag for $ZH \rightarrow q\bar{q}b\bar{b}$ at 250GeV

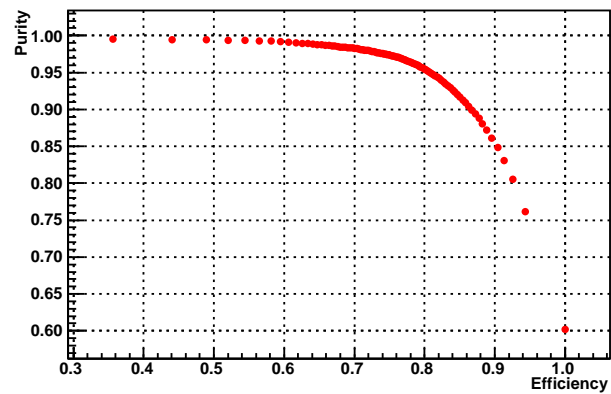
B tag likelihood



Efficiency × Purity



Purity vs. Efficiency



neural network: used variables

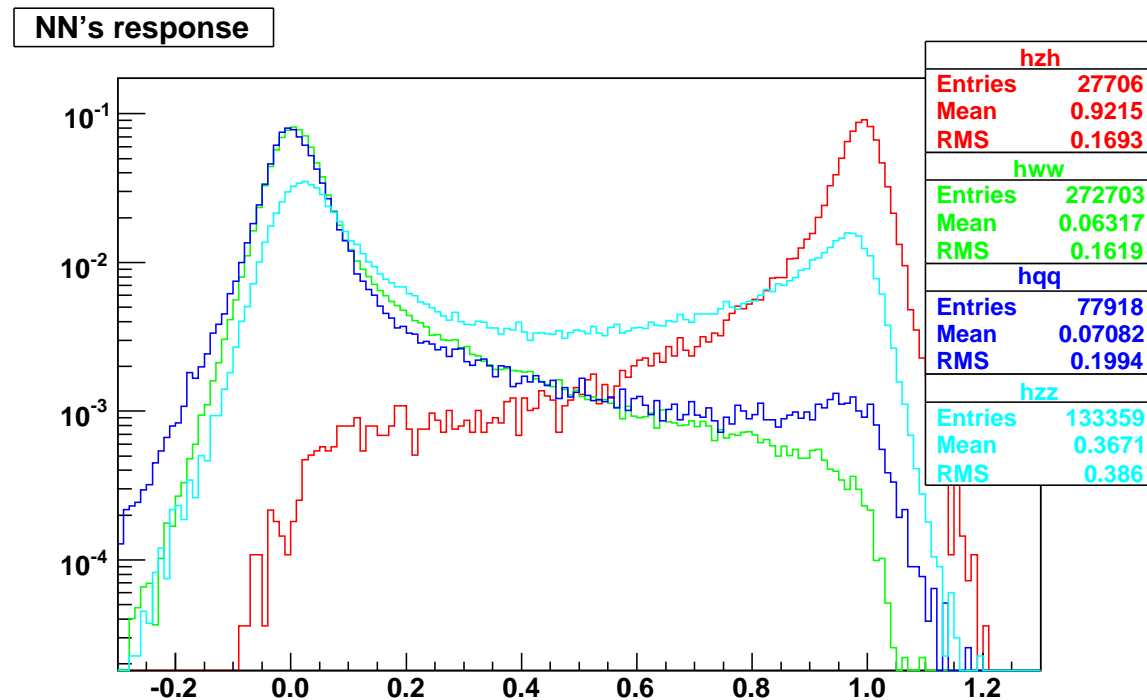
- total visible energy; PFA number; y_{34} ; minimum jet-jet angle
- thrust; theta of thrust axis; sphericity; aplanarity; Fox-Wolfram moments h_{30} and h_{40}
- χ^2 , χ^2 probability and M_{Z^0} of $4C \otimes M_{j_1 j_2} = 91.20$ GeV fitting;
 χ^2 of $4C \otimes M_{j_1 j_2} = M_{j_3 j_4}$ fitting
- j_{mom} , j_{ang} , modified Nachtmann-Reiter angle @ OPAL CERN-EP/98-167
 - sort jets by jet energy $E_{j_1} \geq E_{j_2} \geq E_{j_3} \geq E_{j_4}$
 - $j_{mom} = \frac{|\vec{p}_1| + |\vec{p}_2| - |\vec{p}_3| - |\vec{p}_4|}{\sqrt{s}}$
 - $j_{ang} = \frac{E_4}{\sqrt{s}} (1 - \cos \theta_{12} \cos \theta_{13} \cos \theta_{23})$
 - $|\cos \theta_{N-R}| = \frac{(\vec{p}_1 - \vec{p}_2) \cdot (\vec{p}_3 - \vec{p}_4)}{|\vec{p}_1 - \vec{p}_2| \cdot |\vec{p}_3 - \vec{p}_4|}$
- maximum b-tag; b-likeness of two jets for Higgs; number of b-jet (b-tag > 0.10)

Neural network architecture

- **TMVA package v3.8.13: Artificial Neural Networks MLP**
- **neural network architecture 20:38:1**
 - **input layers: 20 variables**
 - **one hidden layer: 38 nodes**
 - **output layers: one**
- **learning parameters**
 - **number of training cycles: 800**
 - **learning rate: 0.02 (default)**
 - **decay rate: 0.01 (default)**
- **neuron activation function: sigmoid (default)**
synapsis function: sum (default)
learning mode: sequential (default)

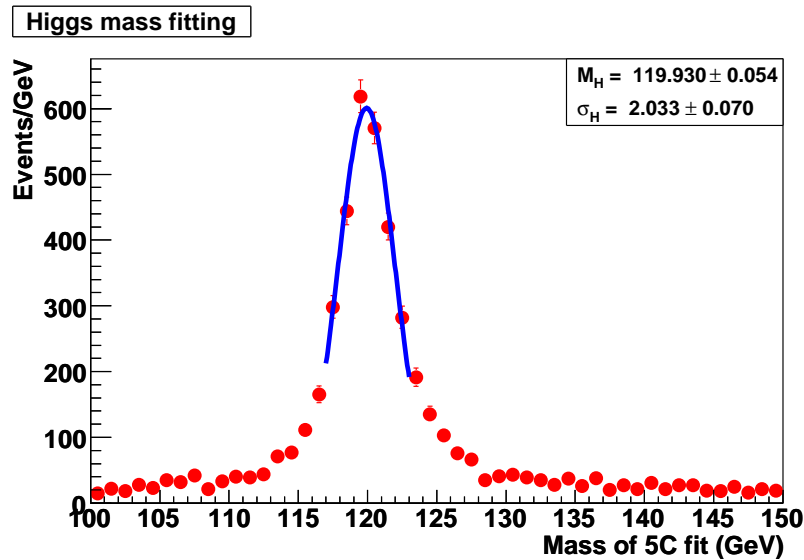
Neural network's response

- the signal peaks around 1.0; the backgrounds peak around 0.0
- plots are normalized to one

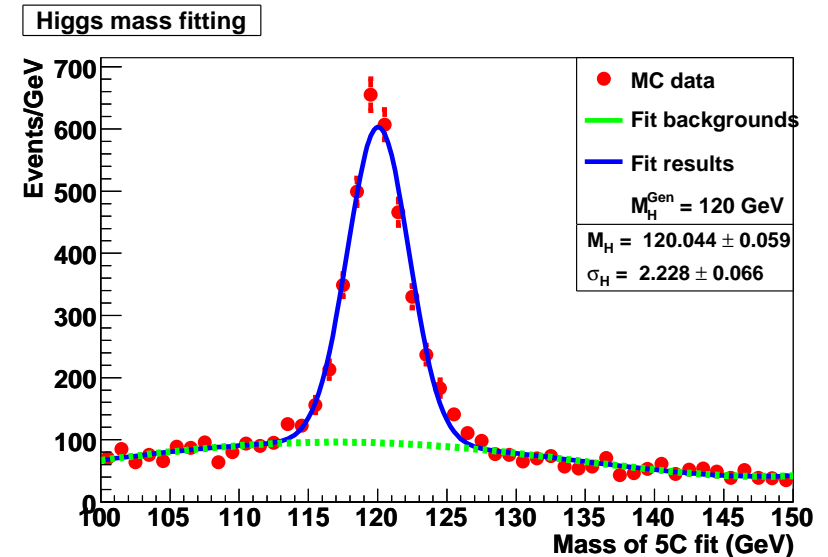


Higgs mass fitting

- only $ZH \rightarrow q\bar{q}b\bar{b}$



- only ZZ background



- 194 WW events @ 300K events \rightarrow 1155 WW @ 250 fb^{-1}
- 179 QQ events @ 500K events \rightarrow 4267 QQ @ 250 fb^{-1}
- need more WW & QQ events for background study

Summary

- $ZH \rightarrow q\bar{q}b\bar{b}$ @ 250GeV for LDCPrime_02Sc
 - about 1340K events
- jet energy correction for udsc-jet and b-jet
- $ZH \rightarrow q\bar{q}b\bar{b}$ selection with neural network
- large WW & QQ samples **future**
 - pre-selection at generator level
 - detector simulation and event reconstruction

Neural network architecture

- neural network architectures
 - how many variables at input layer ?
 - how many hidden layers ?
 - * one hidden layer is sufficient in principle
 - how many nodes at hidden layers ?
 - one output node
- compare different neural network architectures
 - $S/\sqrt{S+B}$ with "0.5-criterion"
 - * Signal (NN's response > 0.5);
 - * Background (NN's response < 0.5)
 - background-to-Signal ratio B/S with "0.5-criterion"
 - gaussian significance S/\sqrt{B} with "0.5-criterion"
- TMVA package v3.8.13: Artificial Neural Networks MLP

Number of variables at input layer

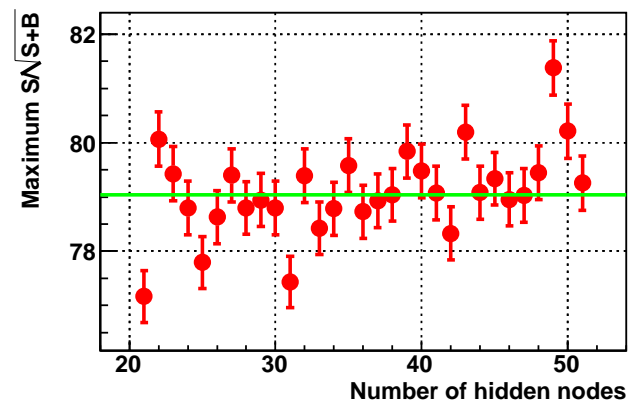
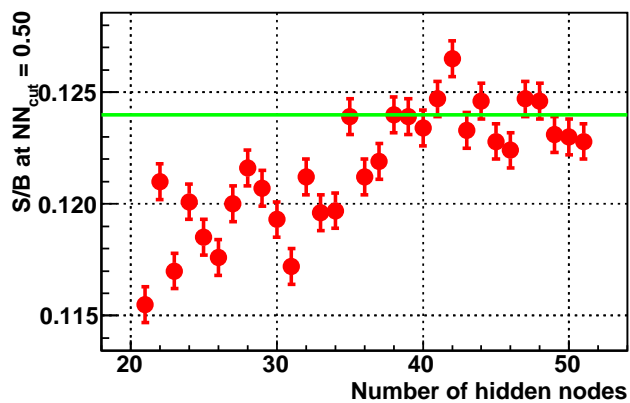
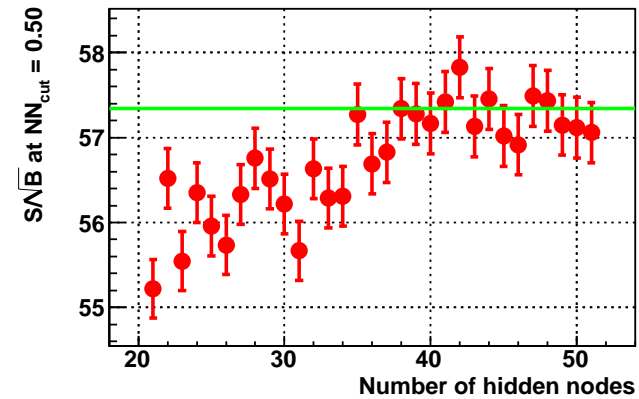
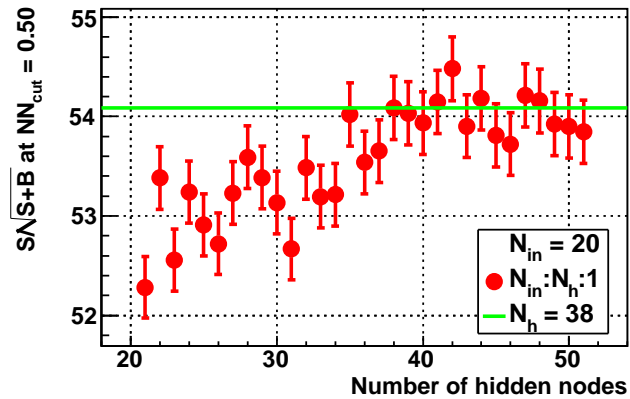
- From 20 variables, remove variable with smallest *Importance* at each step

Importance: sum of the weights-squared of the connections that leave input variable

```
--- MLP      : Ranking result (top variable is best ranked)
--- MLP      : -----
--- MLP      : Rank : Variable      : Importance
--- MLP      : -----
--- MLP      :      1 : aplanarity   : 1.235e+02
--- MLP      :      2 : visible        : 9.100e+01
--- MLP      :      3 : probchi        : 2.375e+01
--- MLP      :      4 : h30            : 2.367e+01
--- MLP      :      5 : jang           : 2.302e+01
--- MLP      :      6 : thrust         : 2.175e+01
--- MLP      :      7 : h40            : 1.890e+01
--- MLP      :      8 : thrusttheta    : 9.918e+00
--- MLP      :      9 : chi2mass       : 6.840e+00
--- MLP      :     10 : ycut           : 6.076e+00
--- MLP      :     11 : sphere         : 3.957e+00
--- MLP      :     12 : blike          : 3.122e+00
--- MLP      :     13 : chi2           : 2.980e+00
--- MLP      :     14 : pfano          : 1.545e+00
--- MLP      :     15 : maxibs         : 1.298e+00
--- MLP      :     16 : jmom           : 9.103e-01
--- MLP      :     17 : bsno           : 1.441e-01
--- MLP      :     18 : massz0         : 1.231e-01
--- MLP      :     19 : nrang          : 5.080e-02
--- MLP      :     20 : smallangle     : 1.654e-02
--- MLP      : -----
```

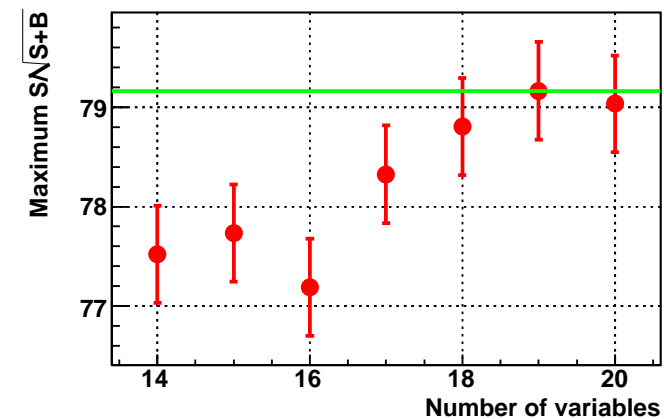
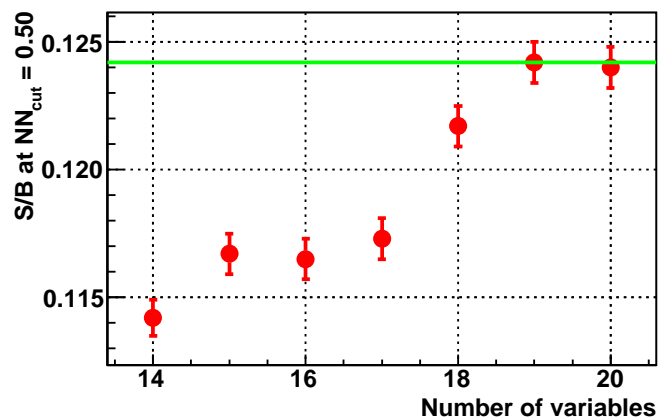
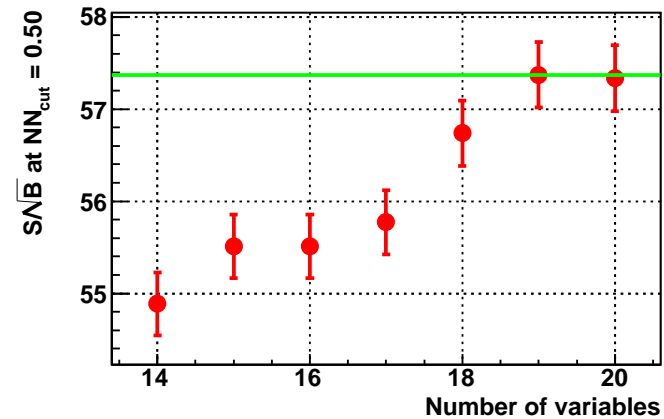
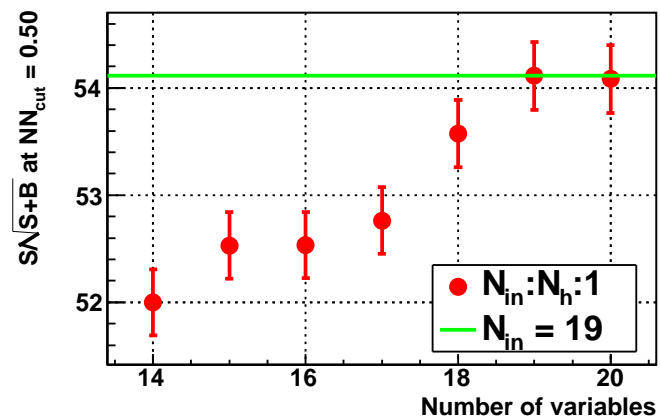
One hidden layer: $N:N_h:1$

- for 20 variables: 20:38:1 ✓



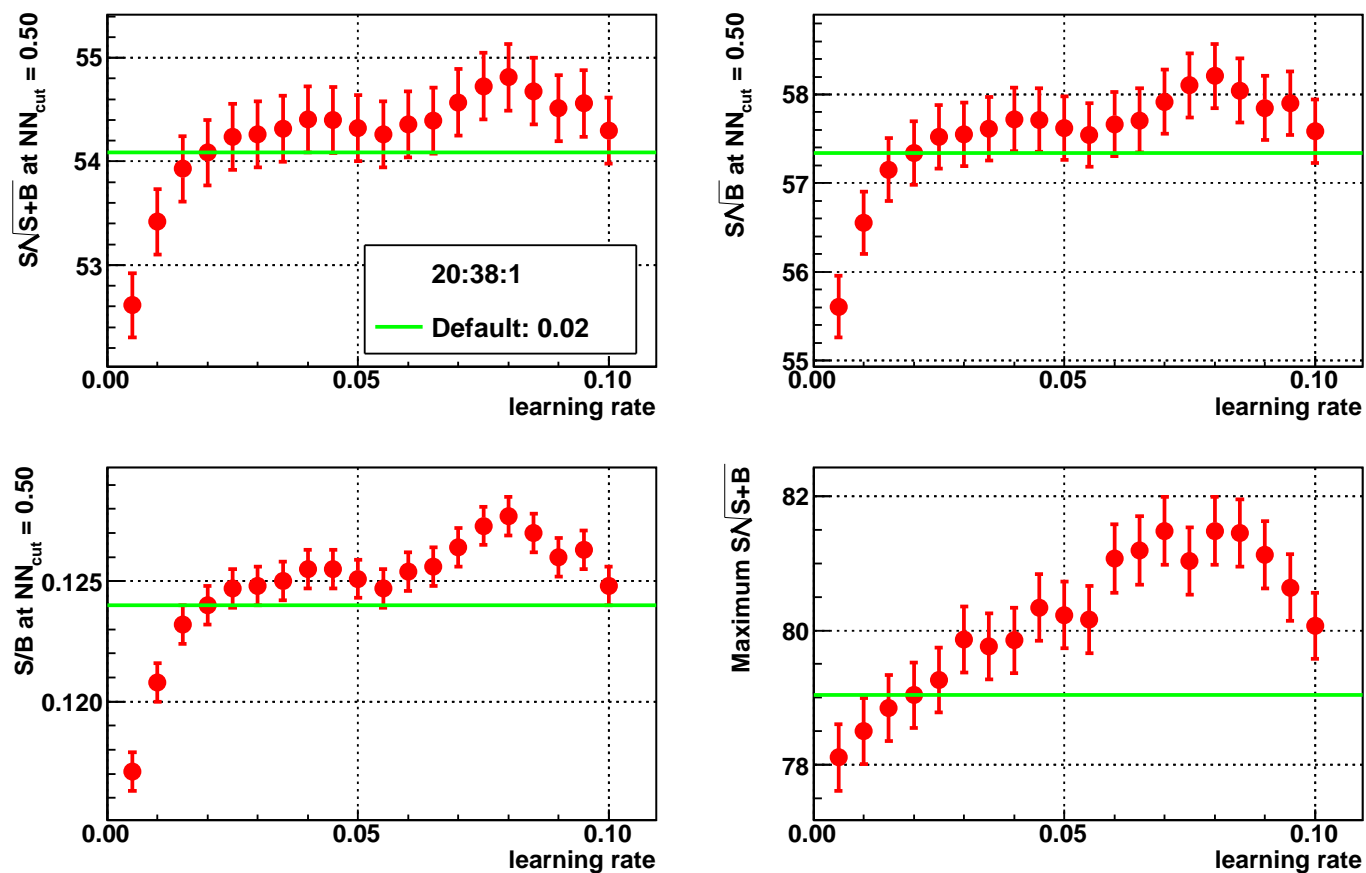
Number of variables at input layer

- good architectures 19:42:1 and 20:38:1 ✓ due to smaller number of weights
19:42:1 has 883 weights; 20:38:1 has 837 weights.



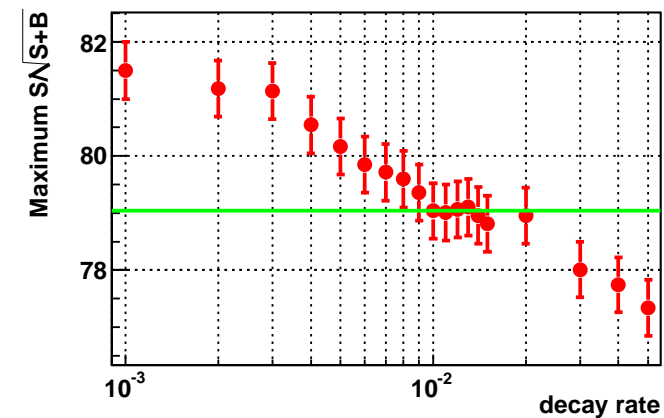
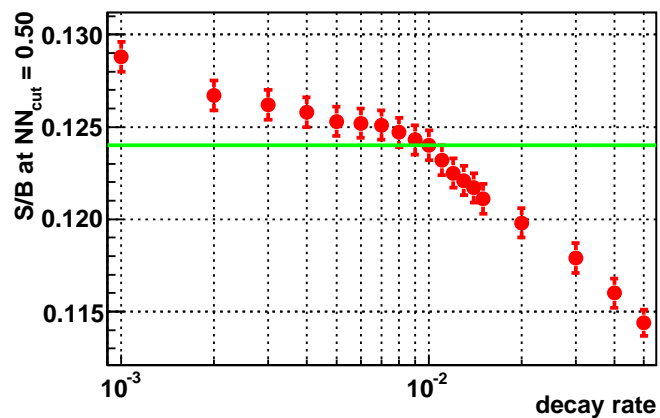
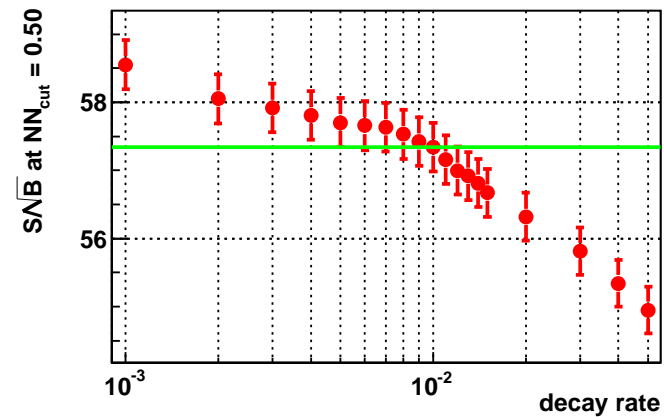
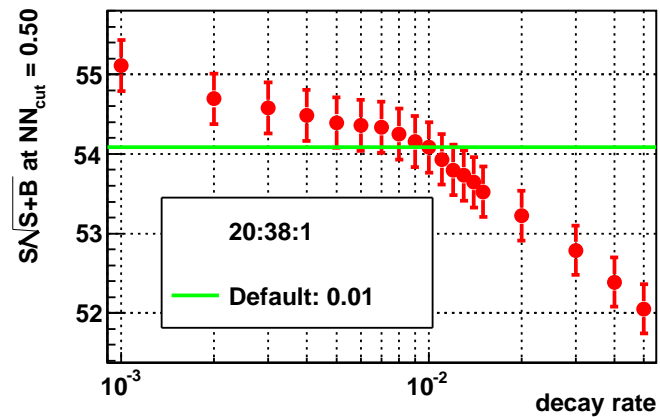
Neural network parameters: learning rate η

- the positive learning rate η is a factor in updating weights



Neural network parameters: decay rate

- the decay rate is a factor for learning parameter; smaller decay rate, larger number of training cycles



Neural network: training sample size

- training sample size: factor \times number of weights

