



Energy reconstruction using MVA in SDHCAL

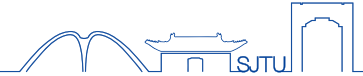
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¹SJTU, ²IPNL July. 02, 2018



上海交通大学
SHANGHAI JIAO TONG UNIVERSITY

Outline



- The introduction of Input and BDT ,MLP structure
 - varList1,2
 - Huge varList
- Energy reconstruction
 - linearity
 - Resolution
- Scikit learn
- Conclusion

Input variables



- **Training events:**
 - 400K 1-120GeV Uniform distributed
- **Application events:**
 - each energy point 20K from 10 to 80GeV

- **Full Input variables:**

nhit1, nhit2, nhit3

nHough, Ncluster, nTrack

nLayer, Density, Radius

nShowerLayer

begin

Huge varList: number of hit1,2,3 of every four layer of SDHCAL , So total 36 variables

BDT structure:

NTree 2000, MaxDepth 5,
learning rate 0.1

MLP structure:

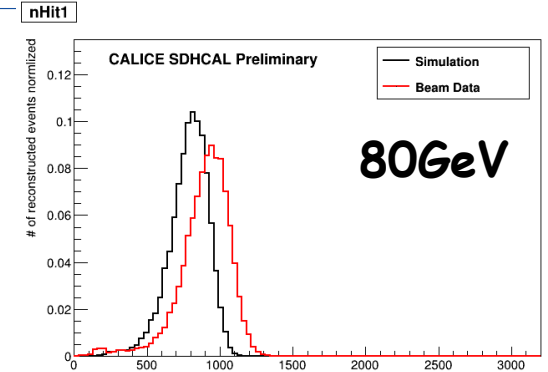
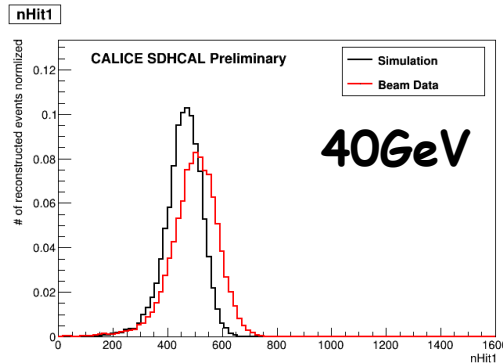
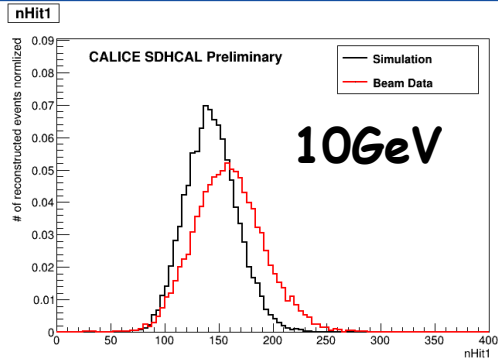
varList1,2: 3 hidden layer with 8,4,2 neurons for corresponding hidden layer

Huge VarList: 5 hidden layer with 6,12,24,12,6 neurons for corresponding hidden layer

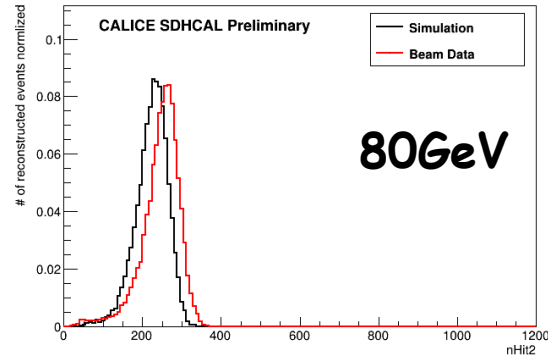
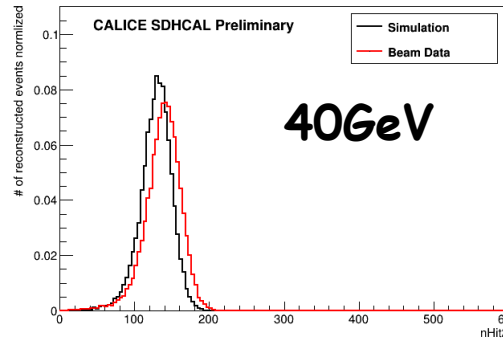
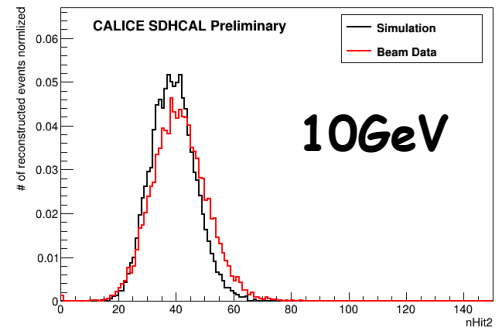
Input Variables



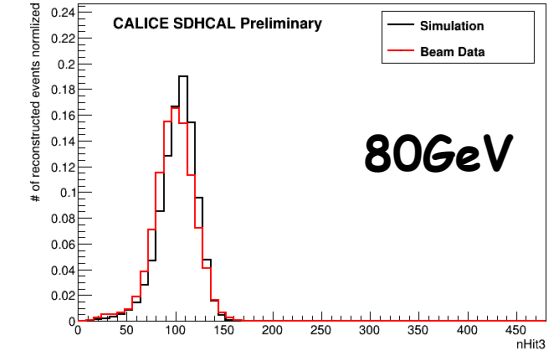
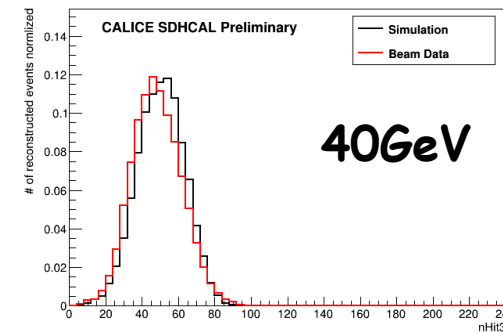
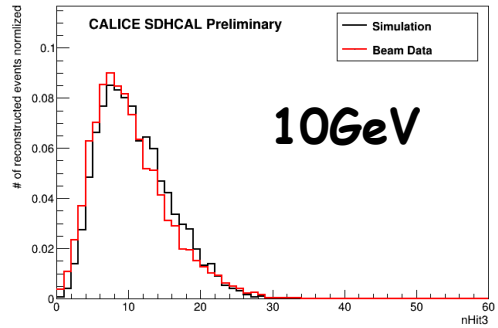
nHit1



nHit2



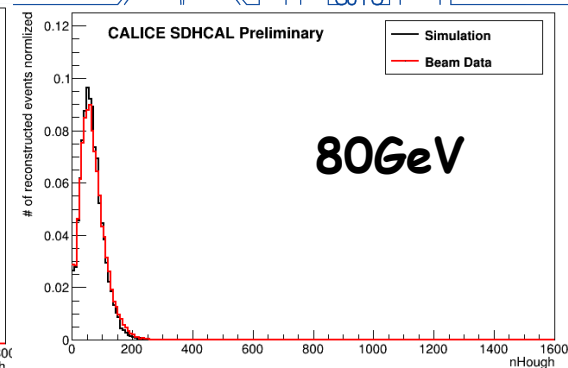
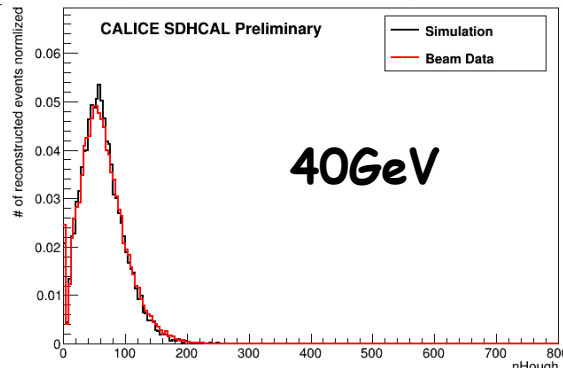
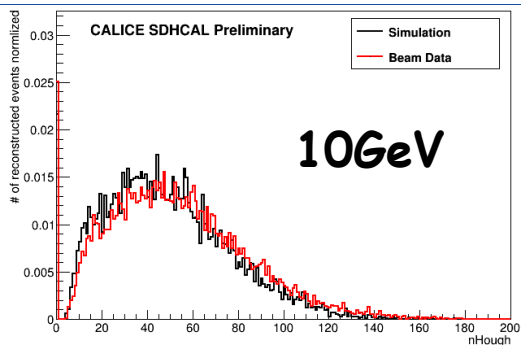
nHit3



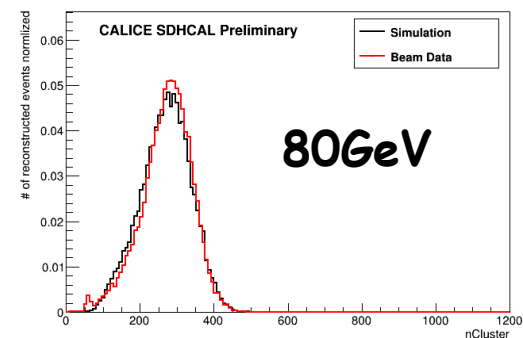
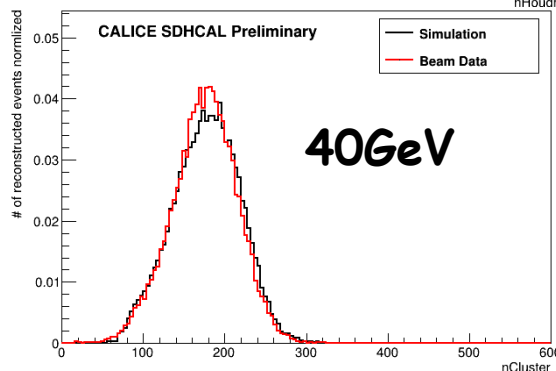
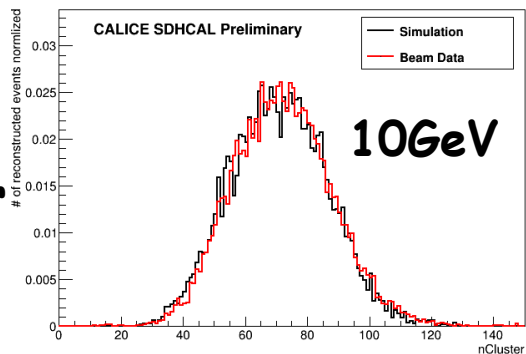
Input Variables



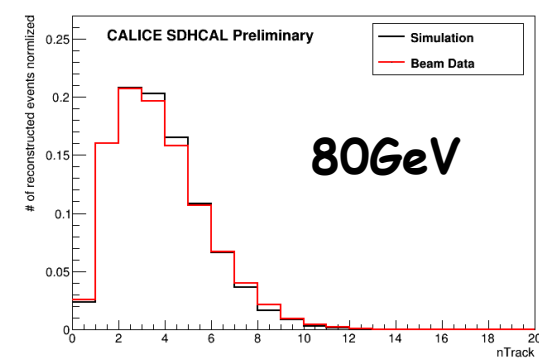
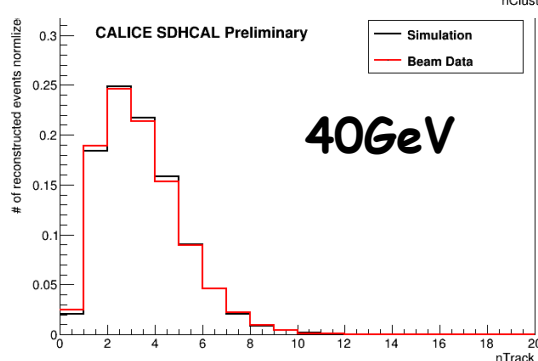
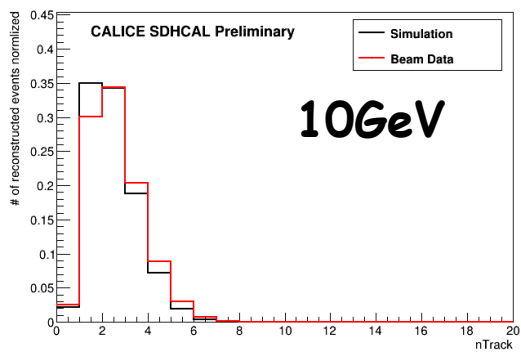
nHough



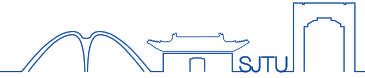
nCluster



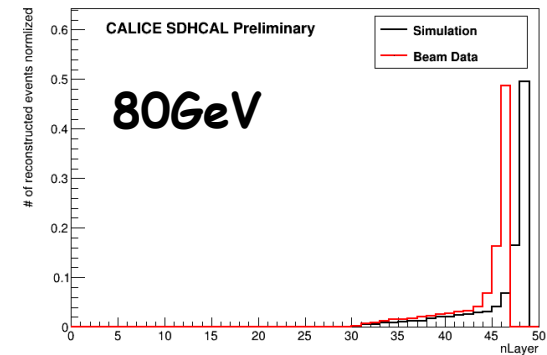
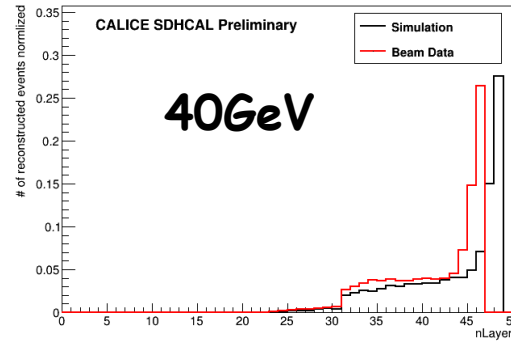
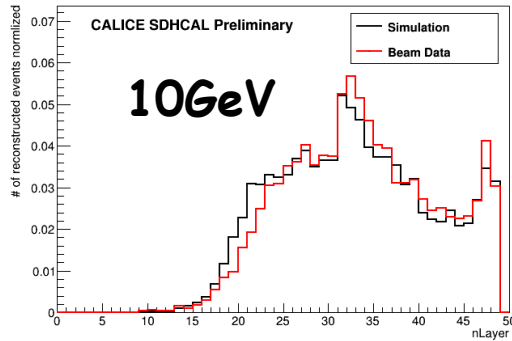
nTrack



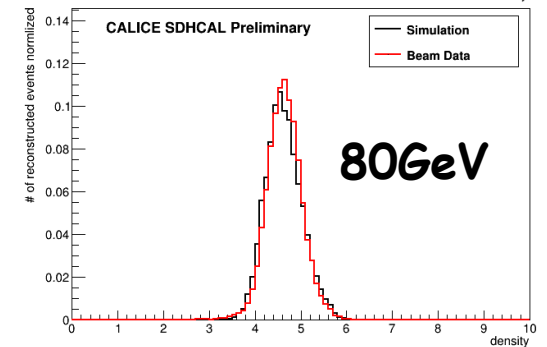
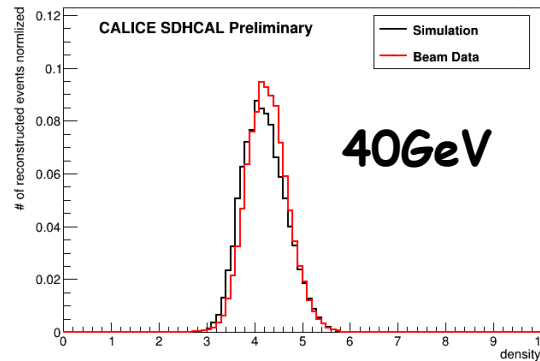
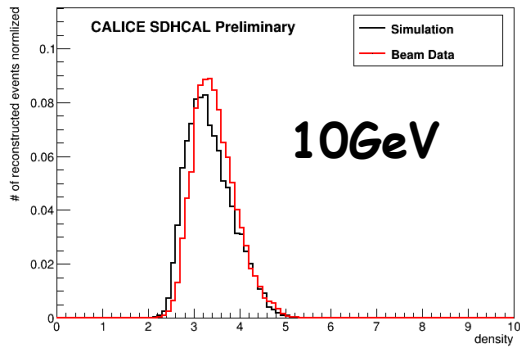
Input Variables



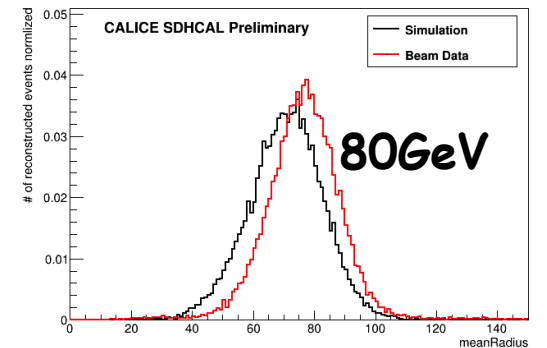
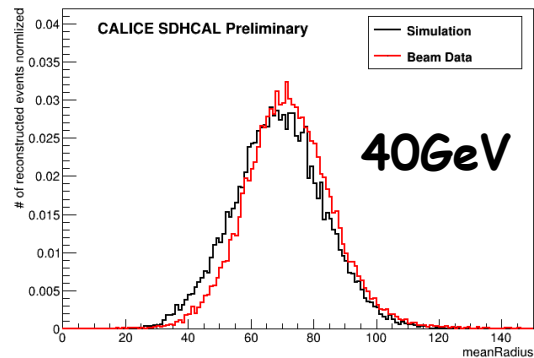
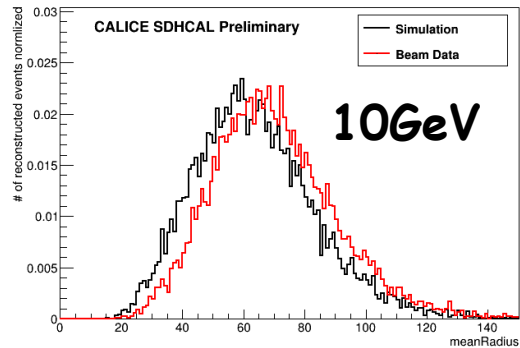
nLayer



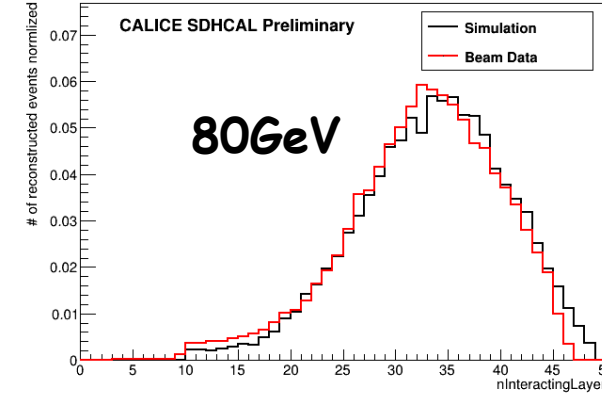
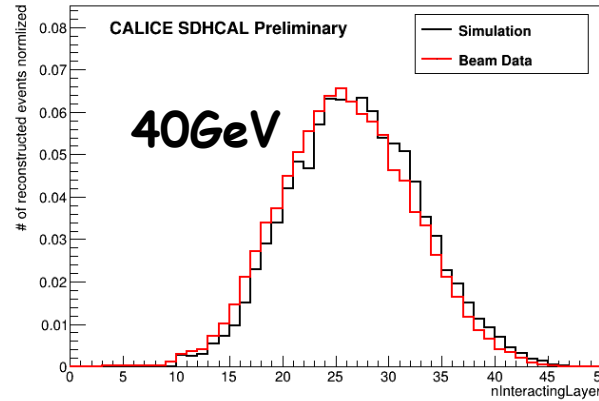
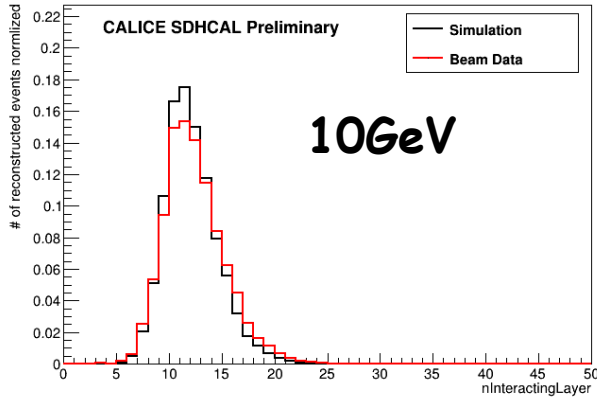
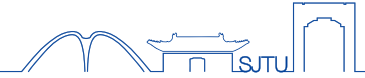
density



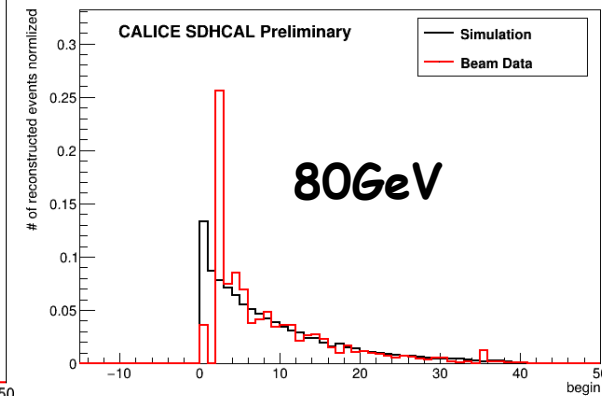
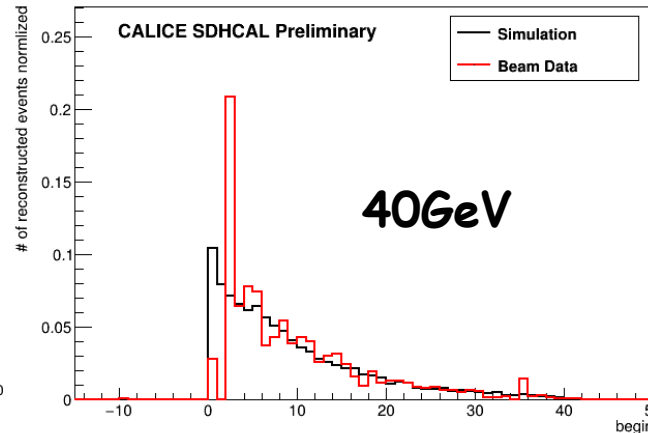
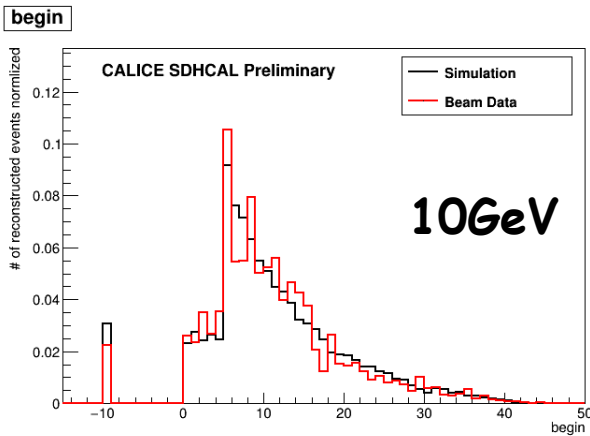
Radius



Input Variables



nShowering Layer

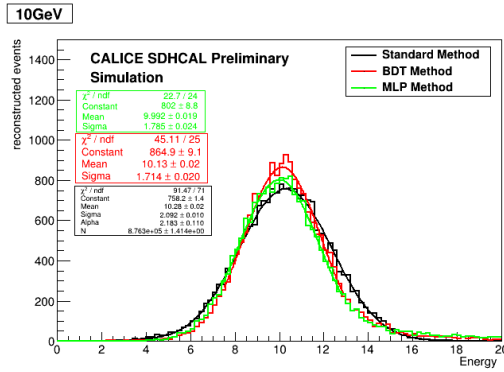


Shower starting layer

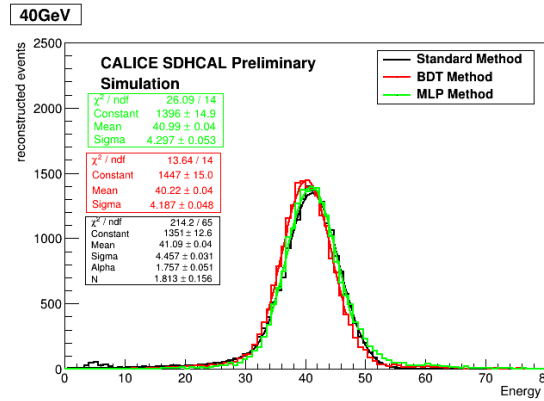


Energy reconstruction

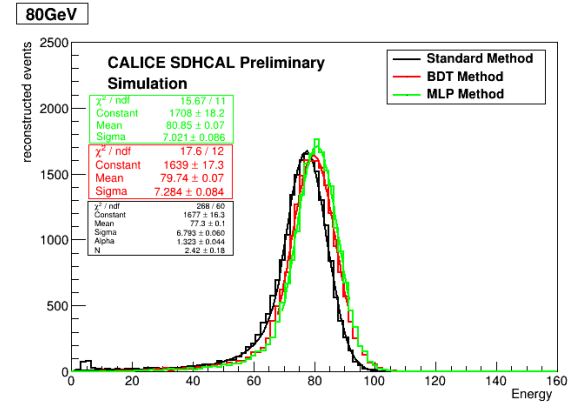
Full VarList training



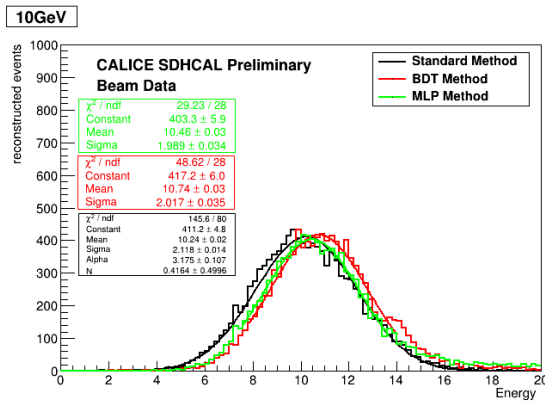
10GeV



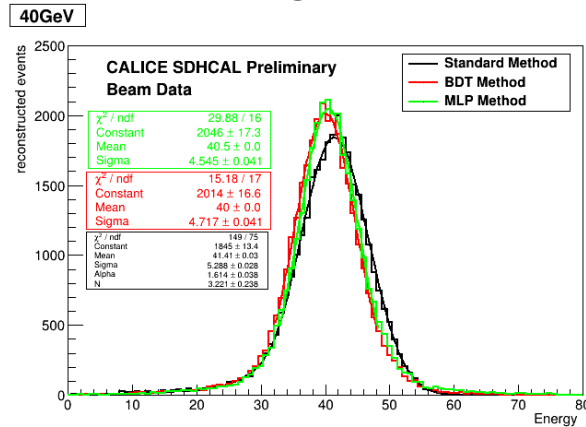
40GeV



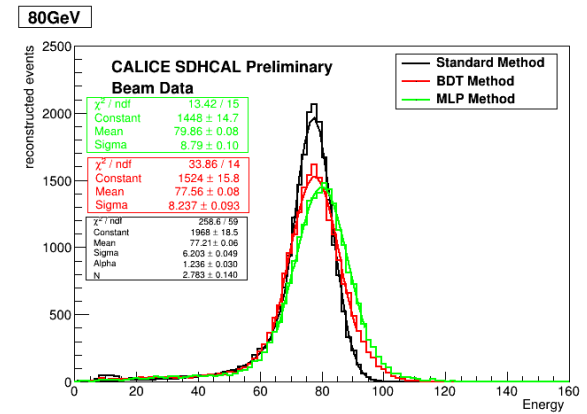
80GeV



10GeV

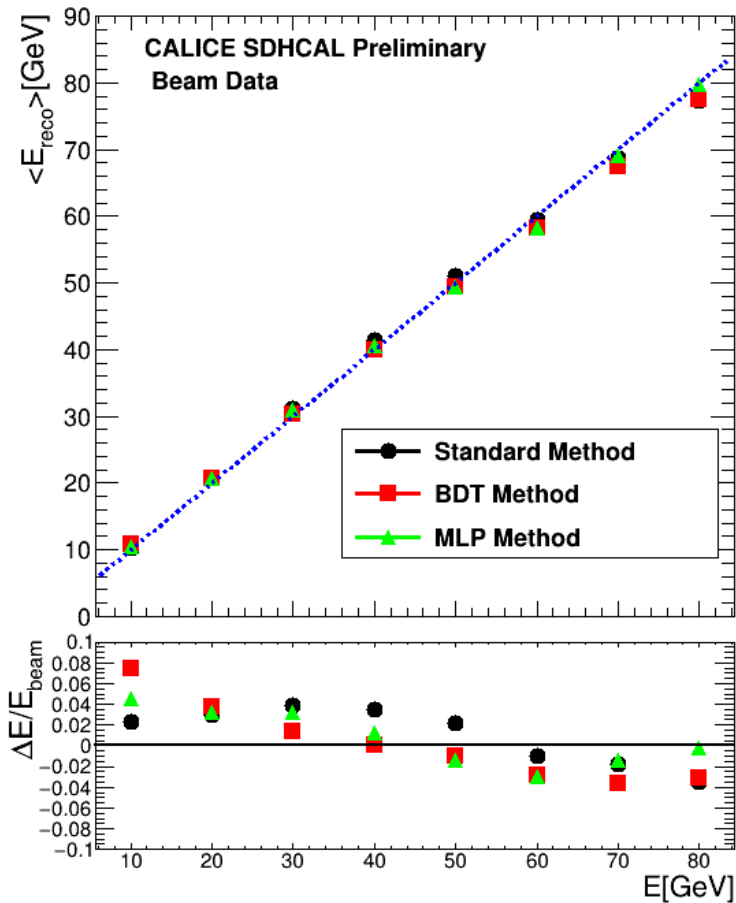
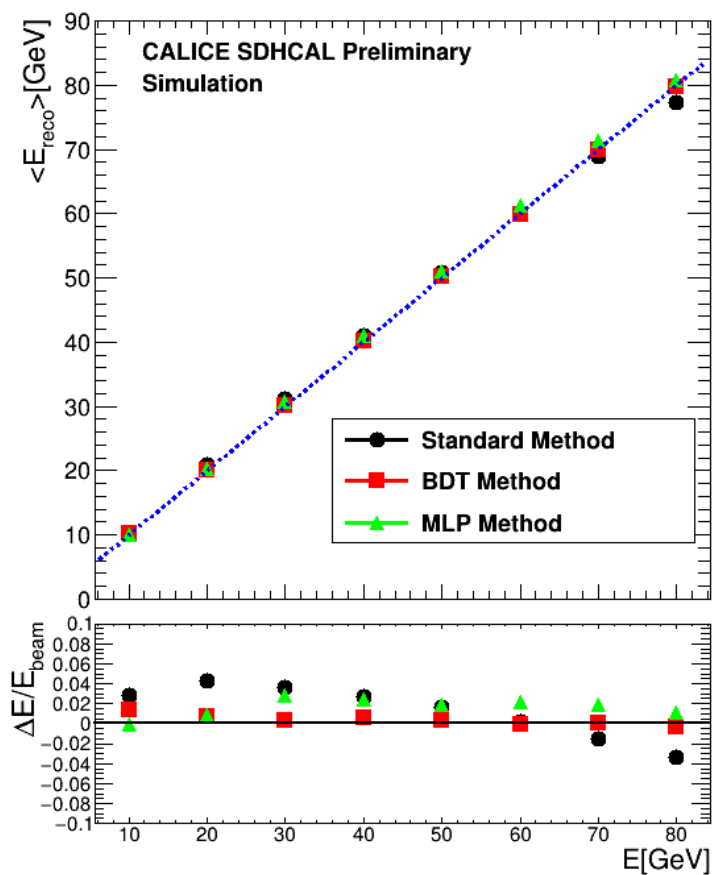


40GeV

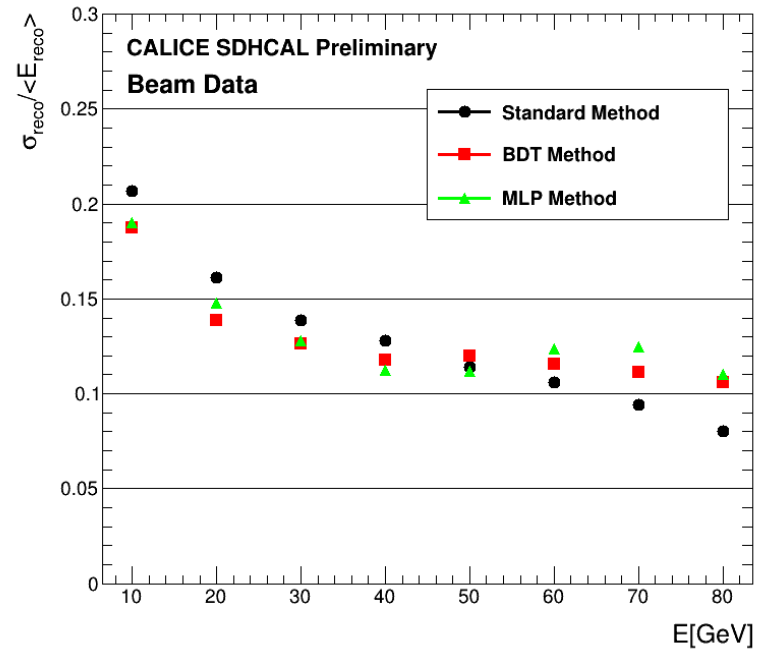
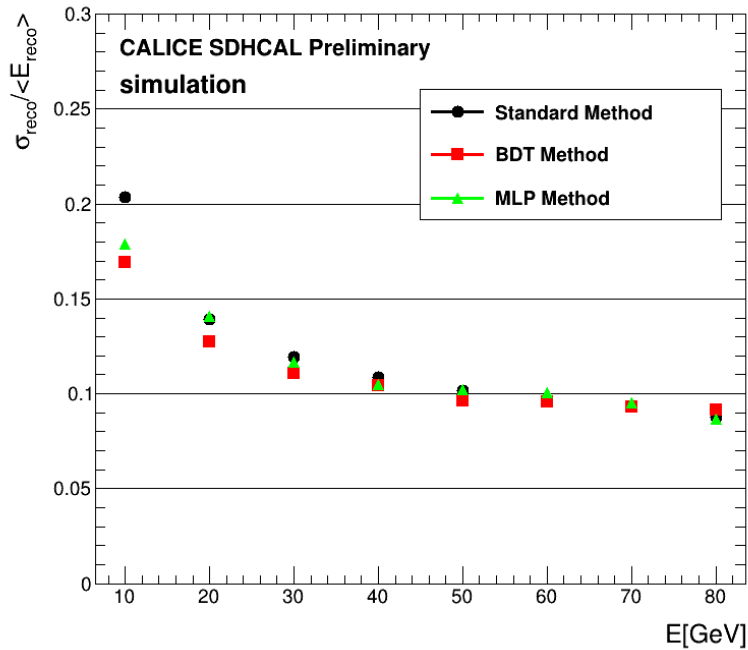


80GeV

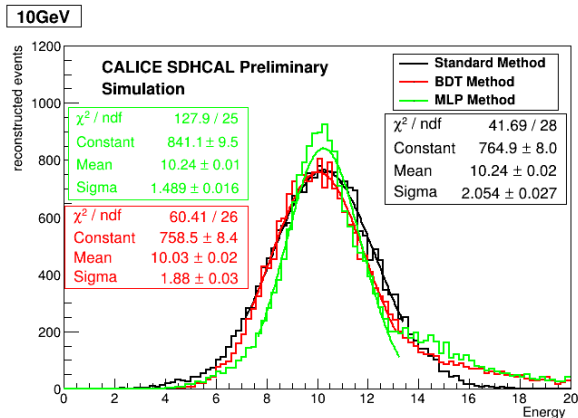
Linearity



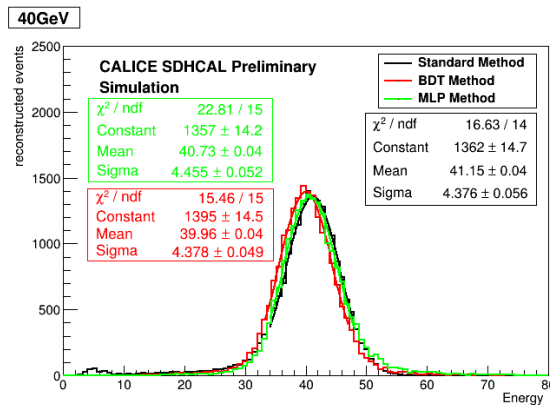
Resolution



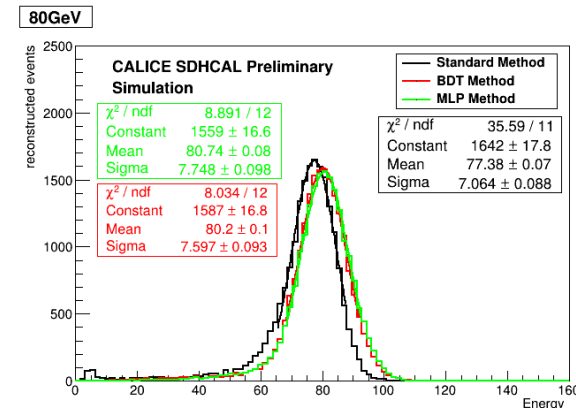
Huge VarList training



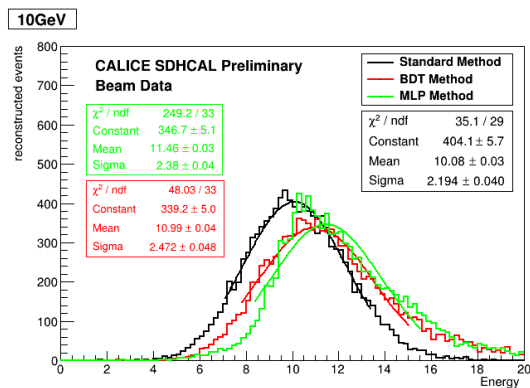
10GeV



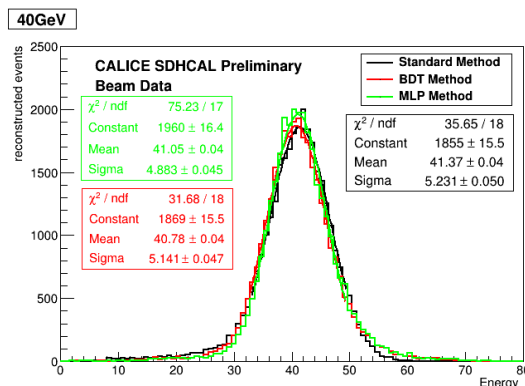
40GeV



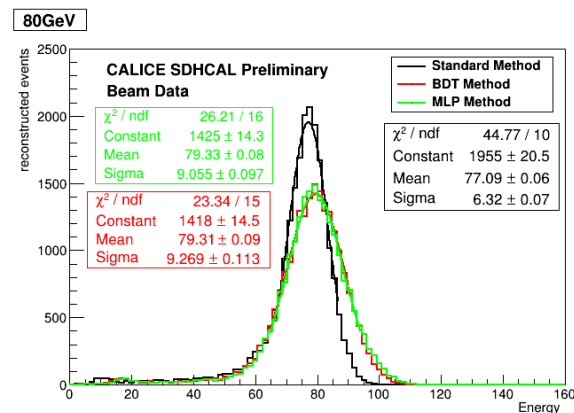
80GeV



10GeV

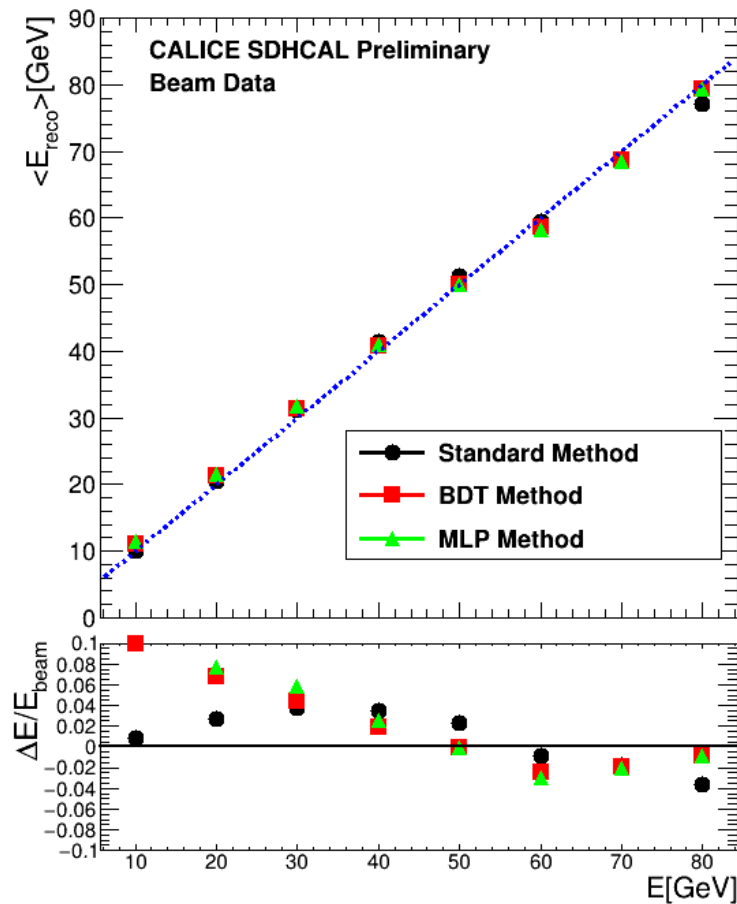
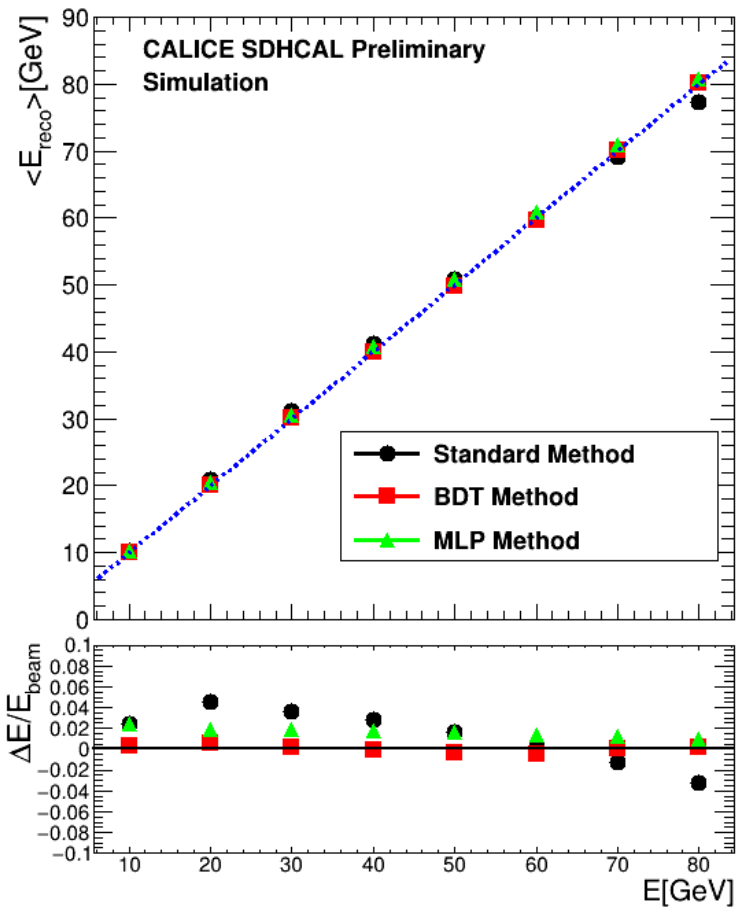


40GeV

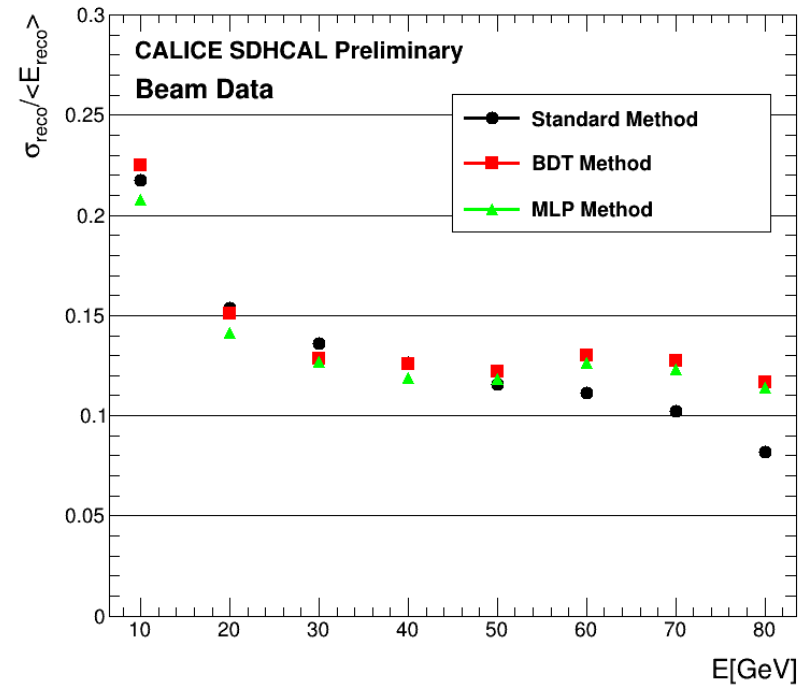
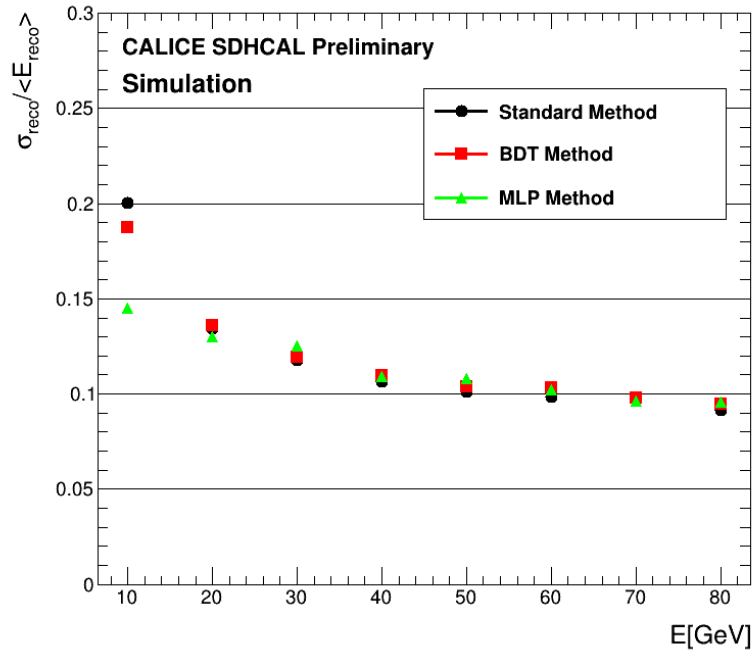


80GeV

Linearity



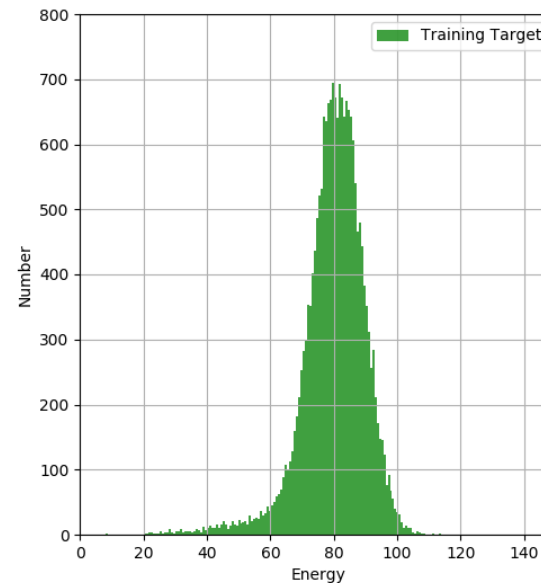
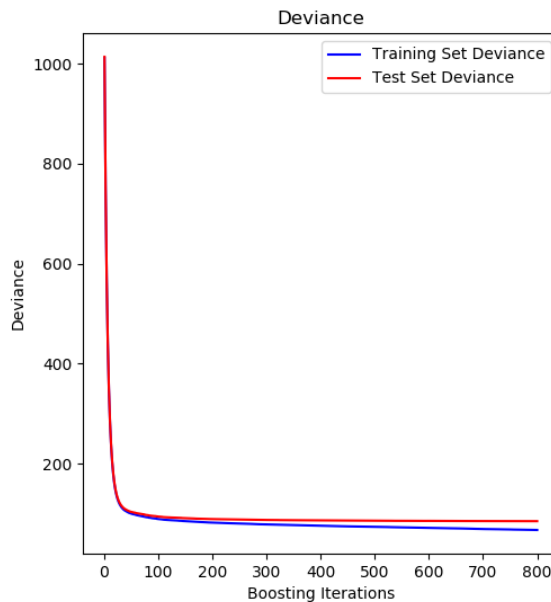
Resolution



Scikit Learn

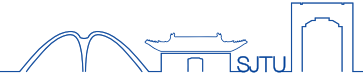


VarList: number of hit1,2,3 of every layer of SDHCAL , So total 144 variables
400K events from 1GeV-120GeV



The analysis is ongoing, roughly it is comparable with TMVA

Conclusion & Next plan



● Energy reconstruction using MVA in SDHCAL

● Simulation

- Good Linearity($\pm 2\%$)
- A improvement of resolution in the low energy

● Data

- It doesn't work well

● Next plan

- Using data samples as training set
- To fine tuning the scikit learn model



Backup

