

Université Catholique de Louvain

Energy Reconstruction and Particle Identification using Neural Network

February 22, 2017

SDHCAL analysis meeting

Outline

- 1 Energy reconstruction: Artificial Neural Network Technique
- 2 Particle identification using MVA in SDHCAL
- 3 Conclusion

Hadronic energy reconstruction in SDHCAL: Artificial Neural Network Technique

Goal: Study of the hadronic energy reconstruction using Neural Network Technique applied in Monte Carlo simulation and beam tests data

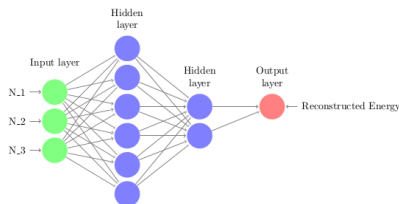
Energy reconstruction: Artificial Neural Network Technique

Simulation Study

- TMultiLayerPerceptron of root package.
- 2 hidden layers with 6 and 2 neurons

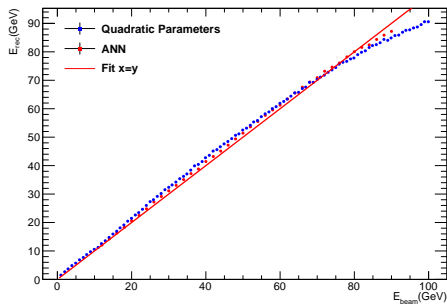
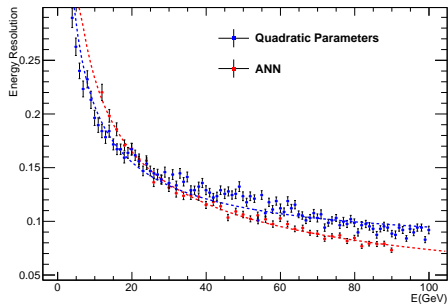
$$\sqrt{\frac{\sum_{i=0}^N (E_{\text{beam}}^i - E_{\text{rec}}^i)^2}{N}}$$

- The input variables: N_1, N_2, N_3 .
- The output variable is the reconstructed energy: E_{rec} .
- Training Samples: Odd energies, 1-99 GeV (50 training samples)
- Test Samples: Even energies, 10-90 GeV (40 test samples)



Energy reconstruction: Artificial Neural Network Technique

Simulation Study

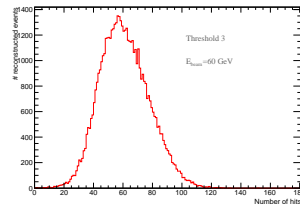
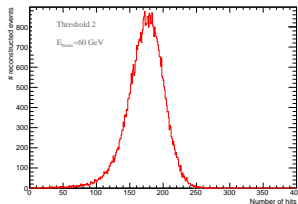
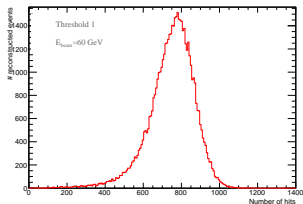
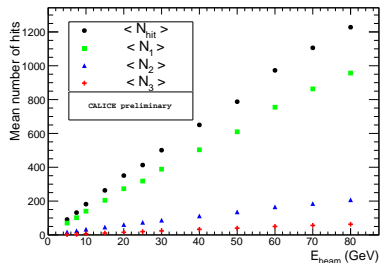


- The Neural Network technique provides a very good improvement in energy resolution at energies higher than 40 GeV.
- Good linearity in the range of $\pm 5\%$.

Energy reconstruction: Artificial Neural Network Technique

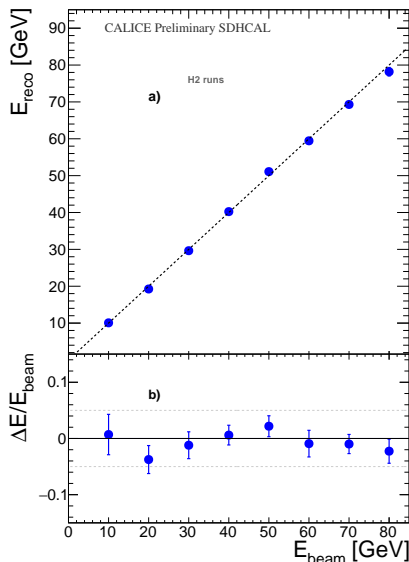
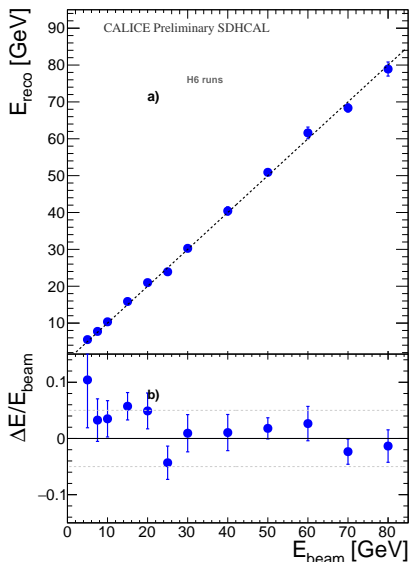
Beam Tests Data

- Architecture of the ANN : One hidden layer of 8 neurons.
- The input variables: N_1, N_2, N_3 .
- The output variable is the reconstructed energy: E_{REC} .
- Data SPS H2 and H6 taken during 2012
 - Training Samples: Trained with Simulation samples, odd energies:1-99GeV
 - Test Samples: 2012 tests beam data: Energies(5,7.5,10,15,20,25,30,40,50,60,70 and 80 GeV)



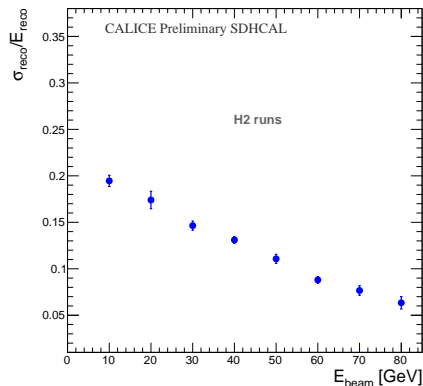
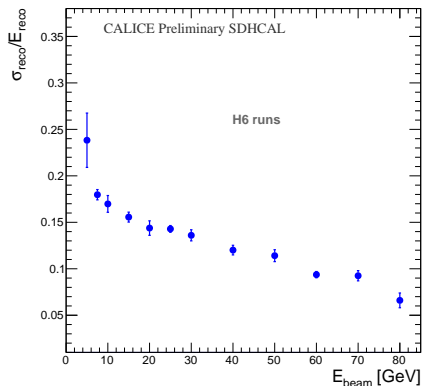
Energy reconstruction: Artificial Neural Network Technique

Beam Tests Data



Energy reconstruction: Artificial Neural Network Technique

Beam Tests Data



The Neural Network technique leads to an improvement in energy resolution reaching 20% at 80 GeV for H2 data compared to the quadratic method.

Particle identification using MVA in SDHCAL

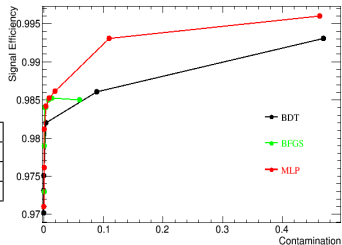
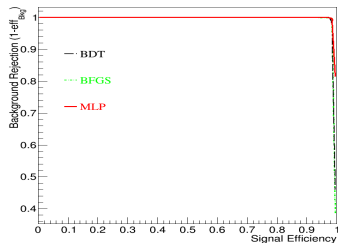
Goal: Development of a particle classifier based on MVA for pion event selection used in the study of the hadronic energy reconstruction in SDHCAL.

Pion identification using MVA in MC simulation

The TMVA methods

- **PID:**
 - Application:
 - Event selection
 - Better estimation in energy reconstruction
 - Separation EM vs Hadron: Shower topology
- Tool: Traditional cuts, MVA.
- TMVA of root, Methods: MLP, BFGS and BDT
- Energy: 80 GeV
 - **Background:** electrons and muons
 - **Signal:** pions

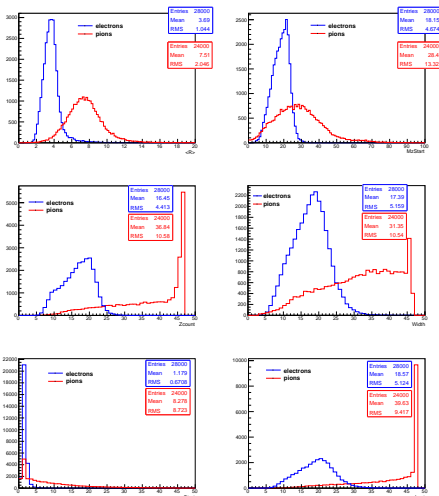
Method	Cut	Efficiency(%)	Contamination(%)	Significance
MLP	0.87	97.1	0.008	31.03
BFGS	0.98	97.3	0.01	31.14
BDT	0.1	97	0.007	31.03



Pion identification using MVA in MC simulation

Performance of MLP method in Pion identification

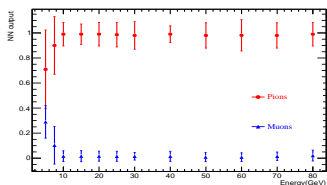
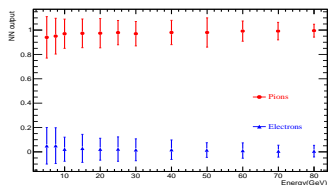
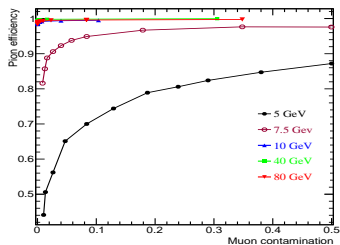
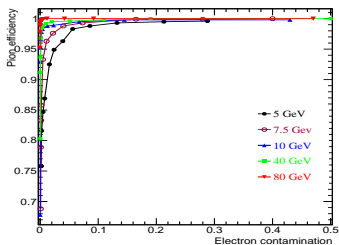
- TMVA of root.
- Method: Multilayer perceptron MLP.
- Training and Test
 - **Background:** 28000 electron events with energies: 2,3,8,10,15,20,25,30,40,50,60,70 and 80 GeV
 - **Background:** 9000 muons events with energies: 0.5, 2 and 10 GeV \implies **Mixed Background**
 - **Signal:** pion events with energies: 5,7.5,10,15,20,25,30,40,50,60,70 and 80 GeV
- Variables: the total number of hits, start, width, mean radius, longitudinal centre of gravity of the shower



Pion identification using MVA in MC simulation

Performance of MLP method in Pion identification

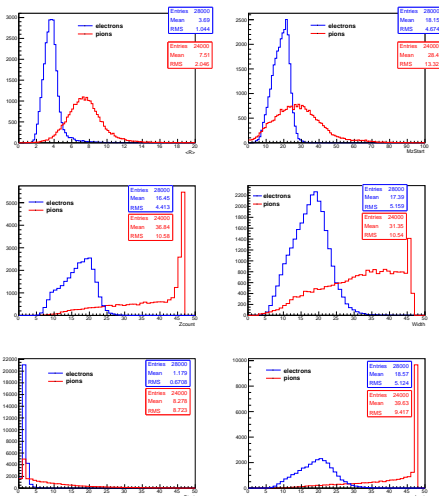
- Good electron/muon separation for the whole energy range
- Good Pion efficiency exceeding 90% for almost the whole energy range with a low electron and muon contaminations ($\approx 1\%$)



Pion identification using MVA in Beam Tests Data

Validation of MLP with beam tests Data

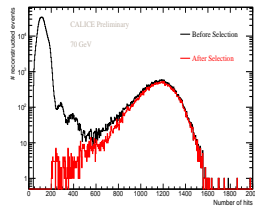
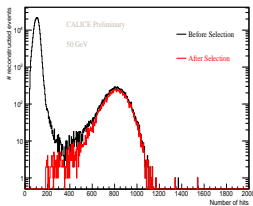
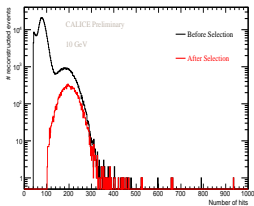
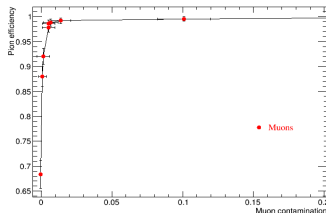
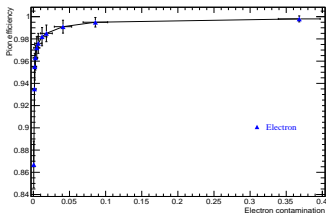
- Method: MLP
- Training and Test
 - **Background:** 28000 electron events with energies: 2,3,8,10,15,20,25,30,40,50,60,70 and 80 GeV
 - **Background:** 9000 muons events with energies: 0.5, 2 and 10 GeV
⇒ **Mixed Background**
 - **Signal:** 24000 pion events with energies(**mixed**): 5,7.5,10,15,20,25,30,40,50,60,70 and 80 GeV
- **Validation:** H2 and H6 beam tests data.
- Variables: the total number of hits, start, width, mean radius, longitudinal centre of gravity of the shower



Pion identification using MVA in Beam Tests Data

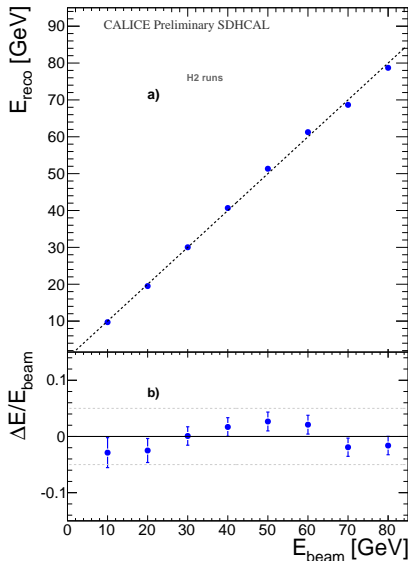
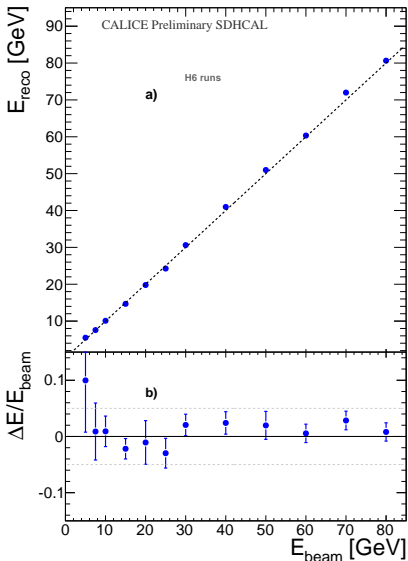
Validation of MLP with beam tests Data

- Good Pion efficiency of 98% for the whole energy range with an electron and muon contaminations of $\approx 1\%$
- Results of MVA event selection is in agreement with the classic event selection



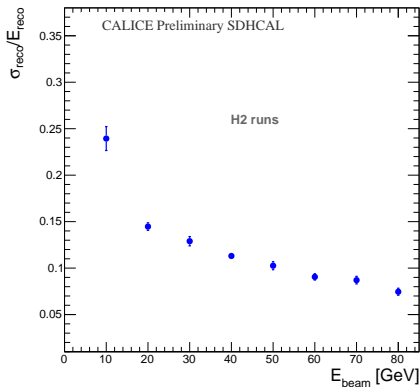
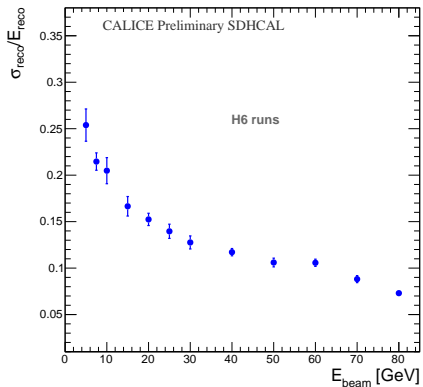
Pion identification using MVA in Beam Tests Data: Energy reconstruction

Quadratic Parametrization



Pion identification using MVA in Beam Tests Data: Energy reconstruction

Quadratic Parametrization



- Good linearity in the range of $\pm 5\%$
- Good energy resolution in agreement with the previous results

Conclusion

Conclusion

- 1 Neural Network has been applied for energy reconstruction in SDHCAL
 - A linear response (within $\pm 5\%$)
 - A significant improvement in energy resolution compared to the quadratic method reaching 20% at 80 GeV
- 2 A particle classifier based on MVA has been developed and applied for pion event selection used after in energy reconstruction
 - Pion identification with MVA is reliable: High pion efficiency with low muon and electron contamination have been obtained
 - MVA based event selection is in agreement with the classic selection
 - Resulted selection exploited in energy reconstruction: A linear response (within $\pm 5\%$) and good energy resolution