

# Energy Calibration with ML

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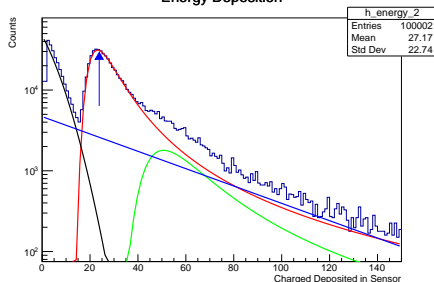
## Energy Calibration

- Sensors in EmCal calibrated by looking for minimum ionizing particles (MIPs)
- Charge deposited by MIPs follows Landau distributions:

$$\text{Signal} = \text{Landau}_0(\mu_0, \sigma_0) + \text{Landau}_1(\mu_1, \sigma_1)$$

- Shown as Red, Green
- Background is composed of a Gaussian pedestal (Black)
  - + high Energy particles (exp, blue), which are signal for analysis, but BG for calibration
- Regression complicated by zero-suppression, which cuts a square notch in a random location between 0 and 10
- Challenge: Fit 200,000 sensors, all with differently shaped signal and background
- Find Most Probable Value ( $\mu$ ) of MIP Landau (Blue Arrow)

Example of charge deposited in single sensor  
Energy Deposition



## Monte Carlo Training and Testing Data Generation

- Training and Testing data generated with Monte Carlo in

generate\_data.C

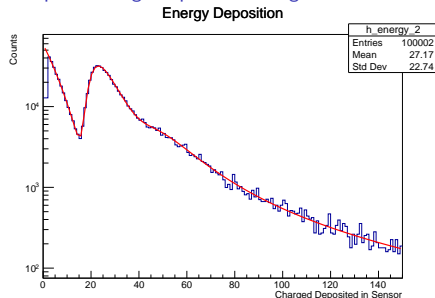
- Signal + Background Function (Red):

Signal =  $\text{Landau}_0(\mu_0, \sigma_0) + \text{Landau}_1(\mu_0, \sigma_0)$

+  $\text{Gauss}(\mu_1, \sigma_1) + \exp(-\tau)$

- Training and testing histograms (blue) generated by randomly sampling MC function 100000 times

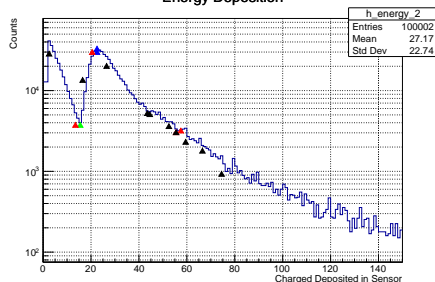
Example of charge deposited in single sensor



## Feature Selection

- Naively fitting using regression with entire underlying functional form fails
- Due to the large number of fit parameters
- Can fit after seeding using features describing histograms shape
- Find following features:
- Local **Minima (Green)** and **Maxima (Blue)**
- Locations where  $\frac{dy}{dx} = 0$  (Red)
- Locations where  $\frac{d^2y}{dx^2} = 0$  (Black)
- Performed in `get_features.C`

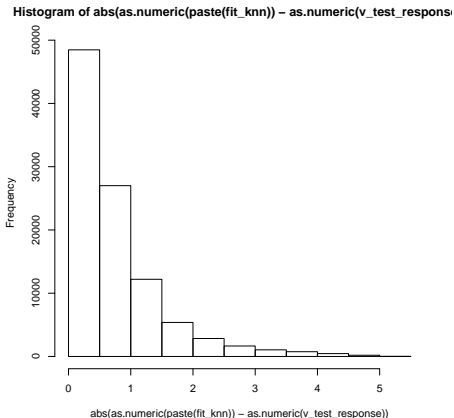
Example of charge deposited in single sensor  
Energy Deposition



## Finding MIP using kNN

- Find MIP location (MPV) using k-Nearest Neighbors
- Features used as predictors
- Predictors: Minima, Mixima,  $\frac{dy}{dx} = 0$
- MIP location (MPV) used as response
- Success measured by distance between true MIP location and predicted MPV
- Fraction of predictions closer that 1 unit from true MIP:
  - For K=1: 75.5%
  - For K=2: 74.4%
  - For K=15: 68.8%
  - For K=100: 60.9%
- Performed in find\_mpv\_KNN.R

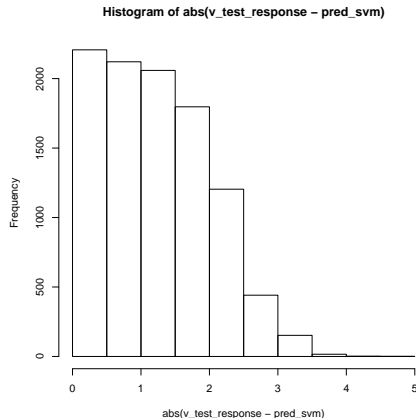
K =1 Difference between true and predicted MIP location (Lower is Better)



K =1 Difference between true and predicted MIP location (Lower is Better)

## Finding MIP using SVM

- Find MIP location (MPV) using Support Vector Machines
- Features used as predictors
- Predictors: Minima, Mixima,  $\frac{dy}{dx} = 0$  and  $\frac{d^2y}{dx^2} = 0$
- MIP location (MPV) used as response
- Success measured by distance between true MIP location and predicted MPV
- Fraction of predictions closer that 1 unit from true MIP:
  - For Radial Kernel with Cost=1: 43.3%



## Finding MIP using a Neural Network

- Finding MIP using a Neural Network
- Uses bin contents of histograms directly instead of feature space
- Input layer: bin contents of histograms
- 2 hidden layers with 10 neurons, 8 neurons
- Output layer: MPV location
- Fraction of predictions closer than 1 unit from true MIP: 69.7%

K =1 Difference between true and predicted MIP location (Lower is Better)

