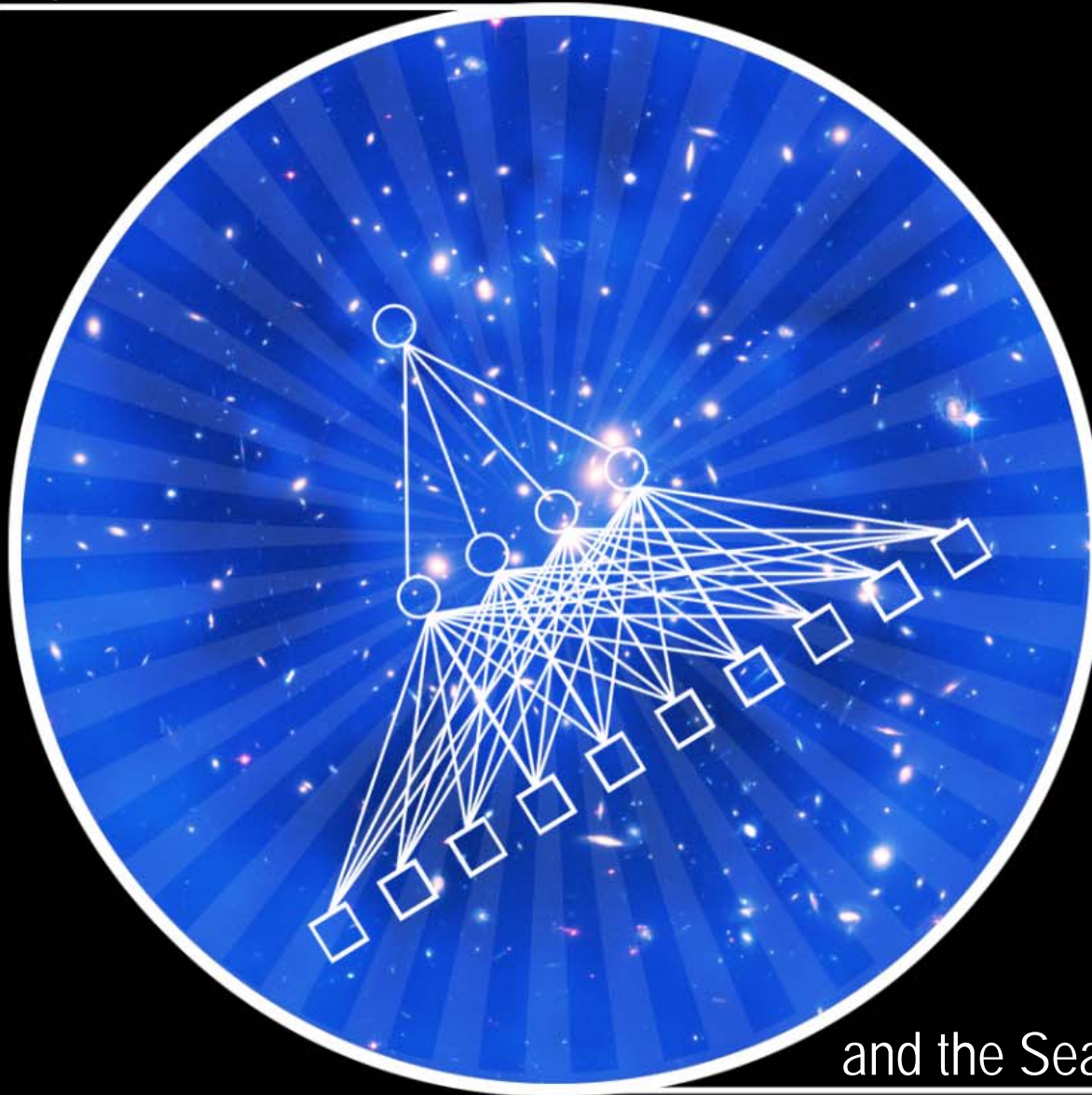


August 2, 2007



Neural Networks
and the Search for Dark Matter

Marc Z. Miskin

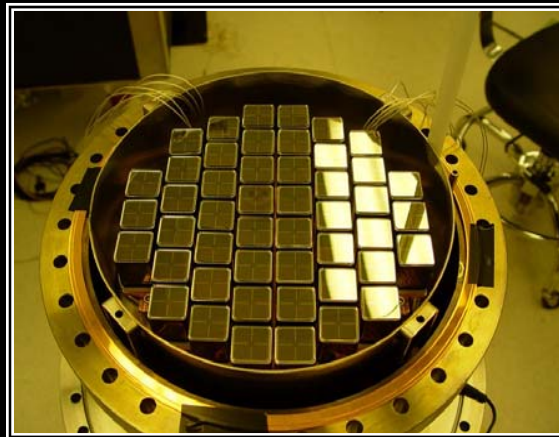
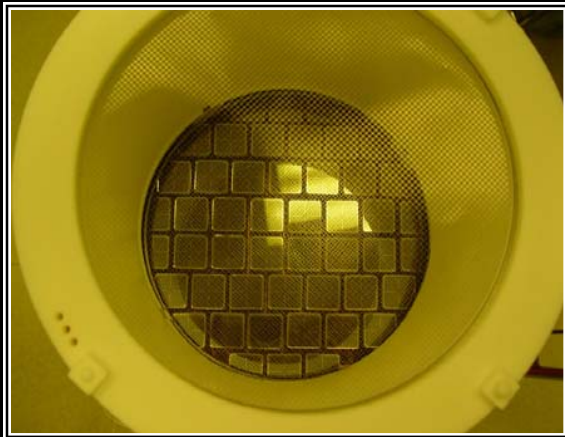
Introduction

- The XENON10 Detector
- Neural Network Overview
- The Double Hit Detection Network



The XENON10 Detector

- The XENON10 Detector uses two arrays of Hamamatsu PMT's to:
 - Convert light to charge and distinguish nuclear recoils from electron recoils
 - Reconstruct events in X and Y to facilitate fiducial volume cuts

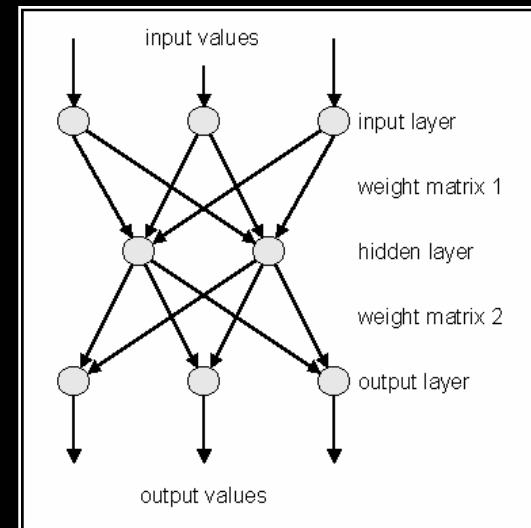


- One source of background which remains is a double hit event, that is a multiple scattering event being recognized as a single event
- Given information from the PMT's, a properly designed NN can recognize and remove such signals

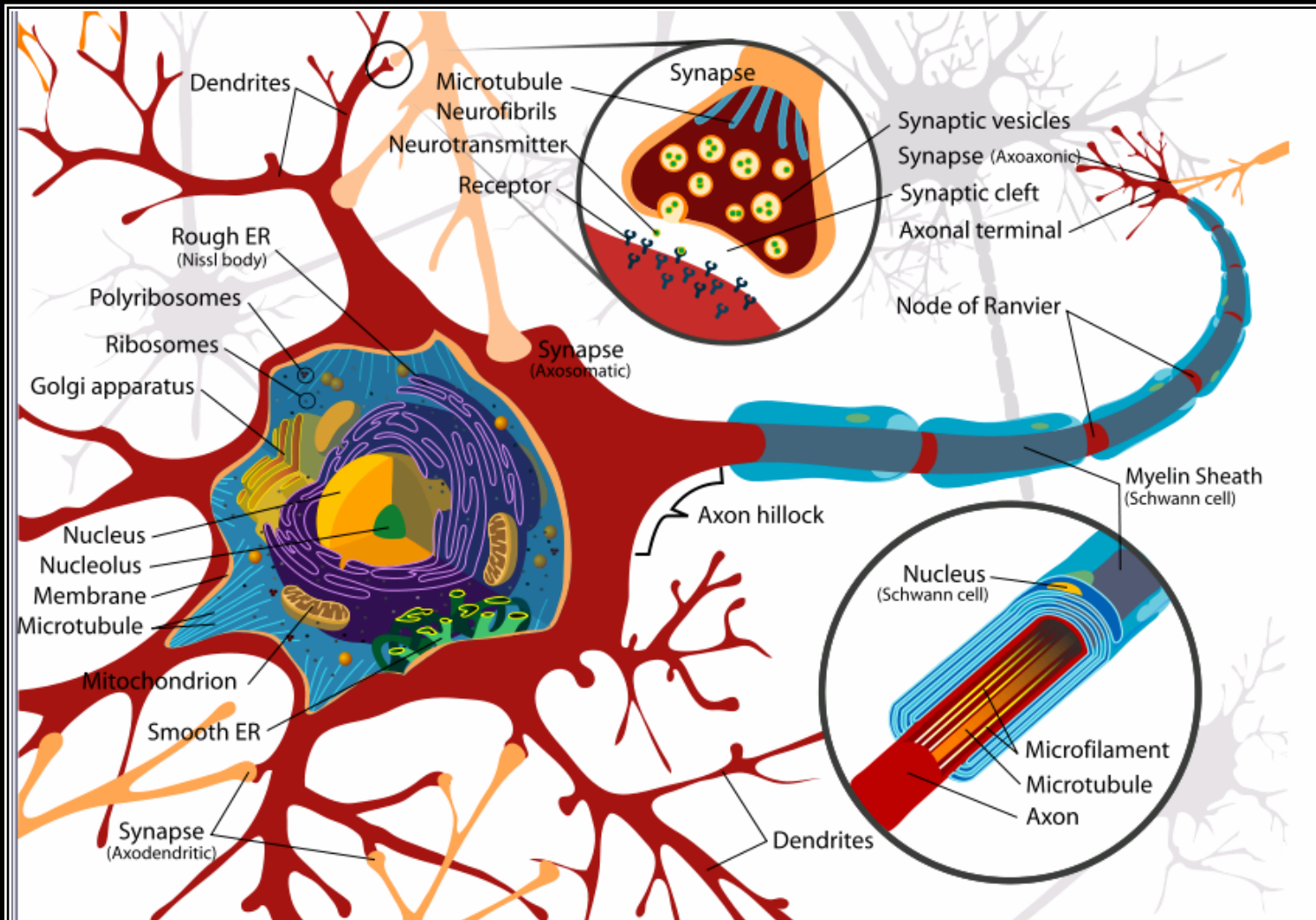


Neural Network

- NN is a processing network modeled on the human brain
 - Processing elements are multiply connected both in and out. (10^{15} - 10^{16} in humans)
 - The inputs to a processing element are weighted and summed (weights can be positive or negative)
 - A threshold is subtracted
 - The result is fed to an activation function (generally a sigmoid) to produce an output
 - Adjusting the weights and thresholds allows the network to produce different responses
 - Given a training set of data, a network can be adjusted automatically to produce proper outputs for certain inputs
 - A well trained network can at times interpolate results well outside its training range



Neural Network



Neural Network

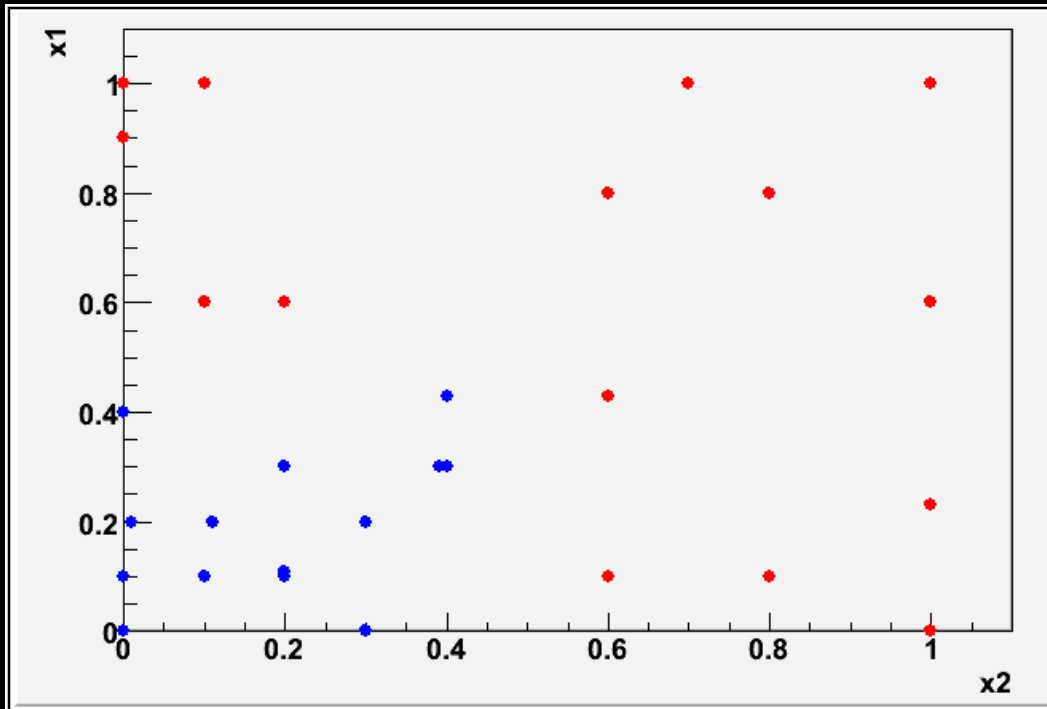


- Learning is analogous to fitting a curve
- There exists some error between the output of the network for a given set of training data and the desired response
- The network weights are adjusted to move across an "error space" and attempt to locate a minimum
- Movement governed by the learning rule
- Repeated for a set number of epochs
- The effectiveness of this mapping is dictated by the topology of the network and the appropriateness of the training set
 - Ex- Pattern complexity
 - Ex- Poor training data
 - Ex- Under fitting and Over fitting

Neural Network



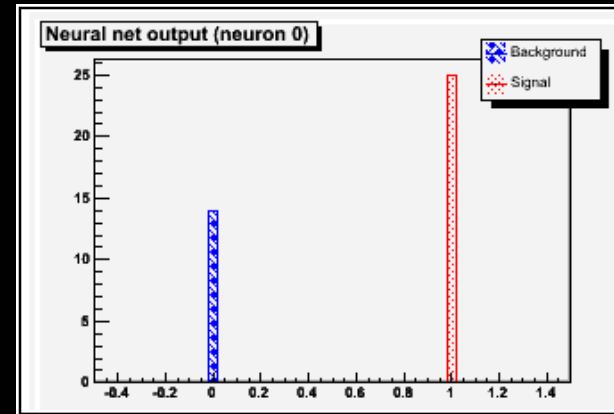
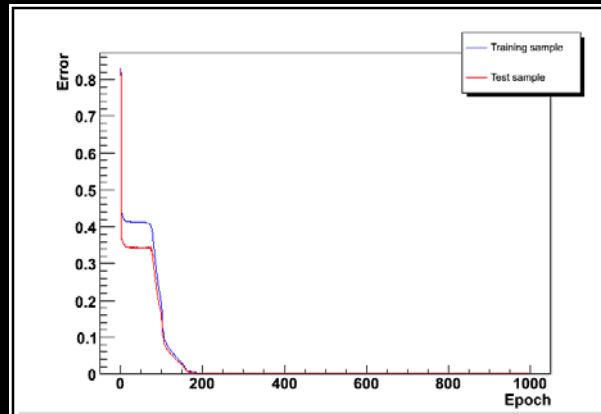
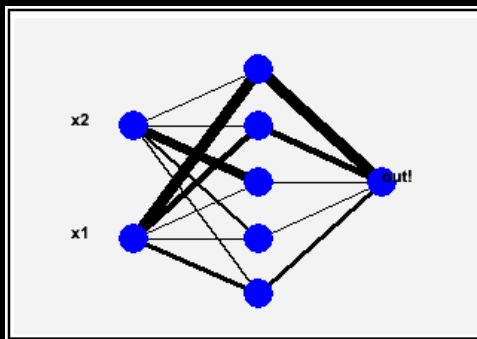
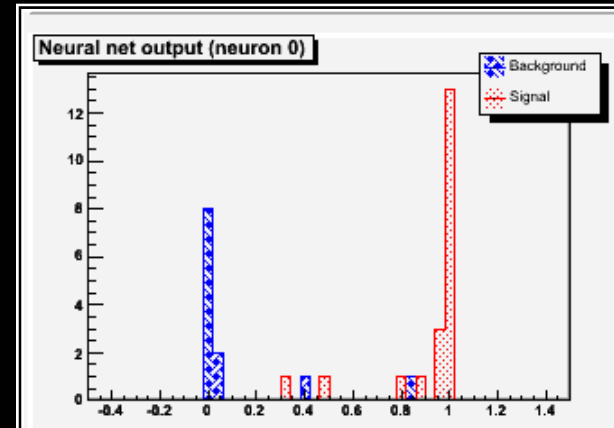
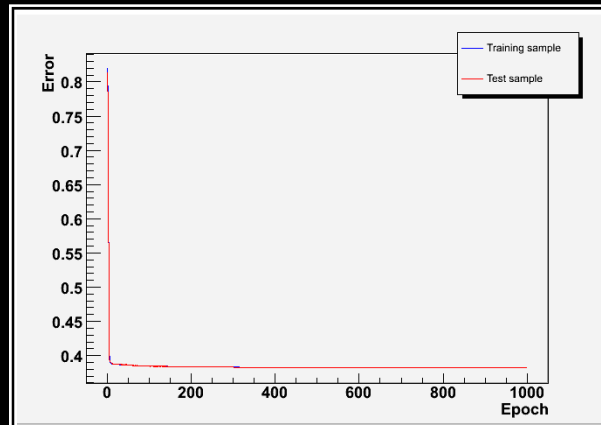
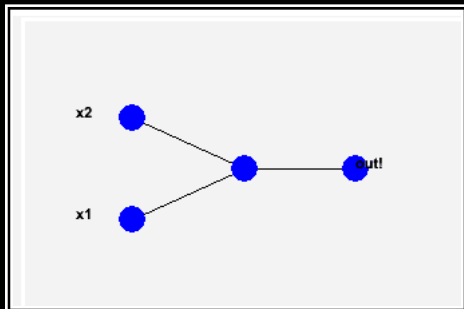
- A basic example: The OR Gate



Neural Network



• Different network designs meet different success



Neural Network



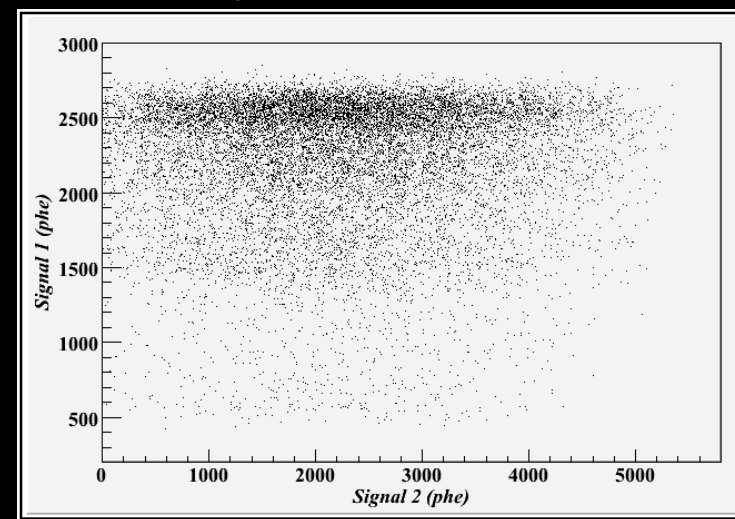
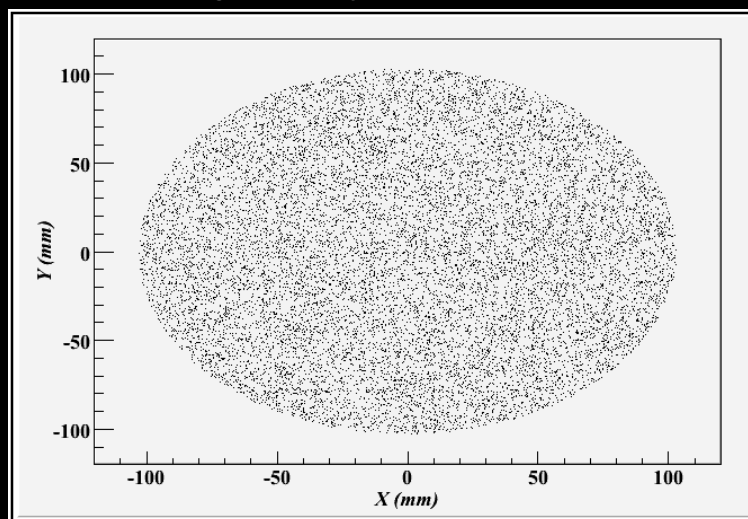
- Comments on the OR Gate:
 - What happens if a negative signal is presented?
 - Note the "saddle" in the training curve
 - What if the training set had been unbalanced?
 - What if the training data were uncorrelated to the modeled pattern?
- Minimizing the error in training is not equivalent to minimizing the actual error.
- Tests must be made to evaluate the true accuracy of the network

The Double Hit Neural Network



- Training data:

- Uses each of the top 47 PMTs as input
- The value at each input is the ratio of the amount of photoelectrons at that PMT to the total quantity of photoelectrons
- 3×10^4 examples randomly distributed in XY
- Equal number of single and double hit examples
- Initial hits generally deposit 2500 phe, the second hit ranges from 0 to 5000 phe



The Double Hit Neural Network



- Network structure
 - Originally designed without a hidden layer, problem proved too complex
 - Changed to use one input layer with 47 neurons, one hidden layer with 47 neurons, and a single neuron output layer
 - Uses hyperbolic tangent activation functions on the input and hidden layers and a sigmoid output function
- Trained until error remained constant for 1000 Epochs

The Double Hit Neural Network



- Networks were tested with 3×10^4 examples randomly distributed in X and Y
- First signal of mostly 25000 phe, second signal in the range of 0 to 5000 phe
- Networks evaluated by MSE, Correlation Coefficient, AIC, MDL

$$MSE = \frac{\sum_{i=0}^{N_{trial}} (d_i - y_i)^2}{N_{trial}}$$

$$AIC = N_{training} \ln(MSE) + 2k$$

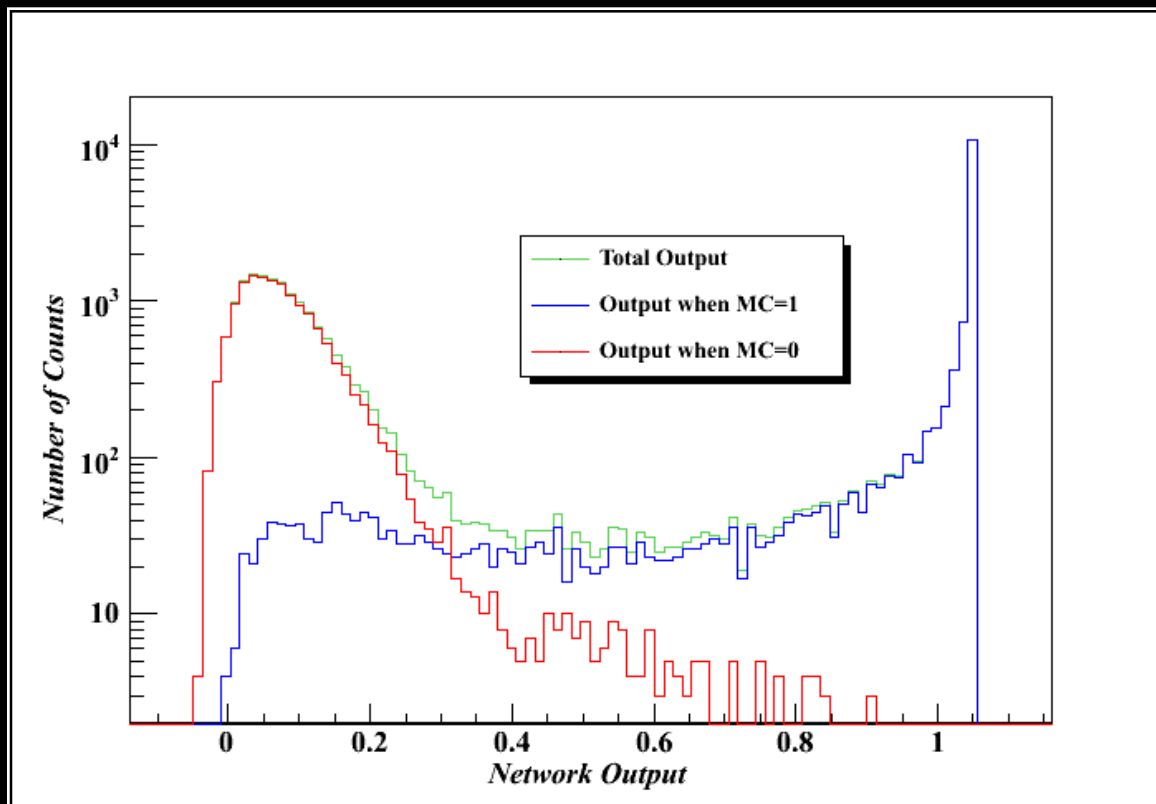
$$MDL = N_{training} \ln(MSE) + 0.5k \ln(N_{training})$$

$$r = \frac{\sum_{i=0}^{N_{trial}} (x_i - \bar{x})(d_i - \bar{d}) / N_{trial}}{\sqrt{\sum_{i=0}^{N_{trial}} (x_i - \bar{x})^2 / N_{trial}} \sqrt{\sum_{i=0}^{N_{trial}} (d_i - \bar{d})^2 / N_{trial}}}$$

The Double Hit Neural Network



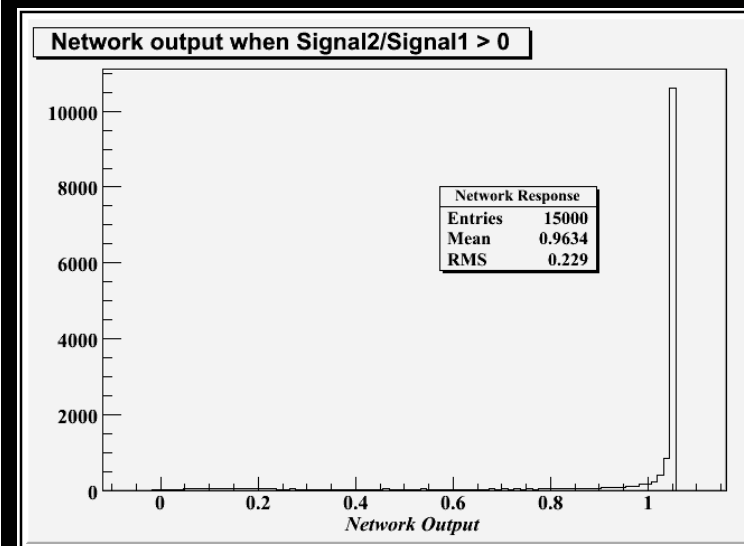
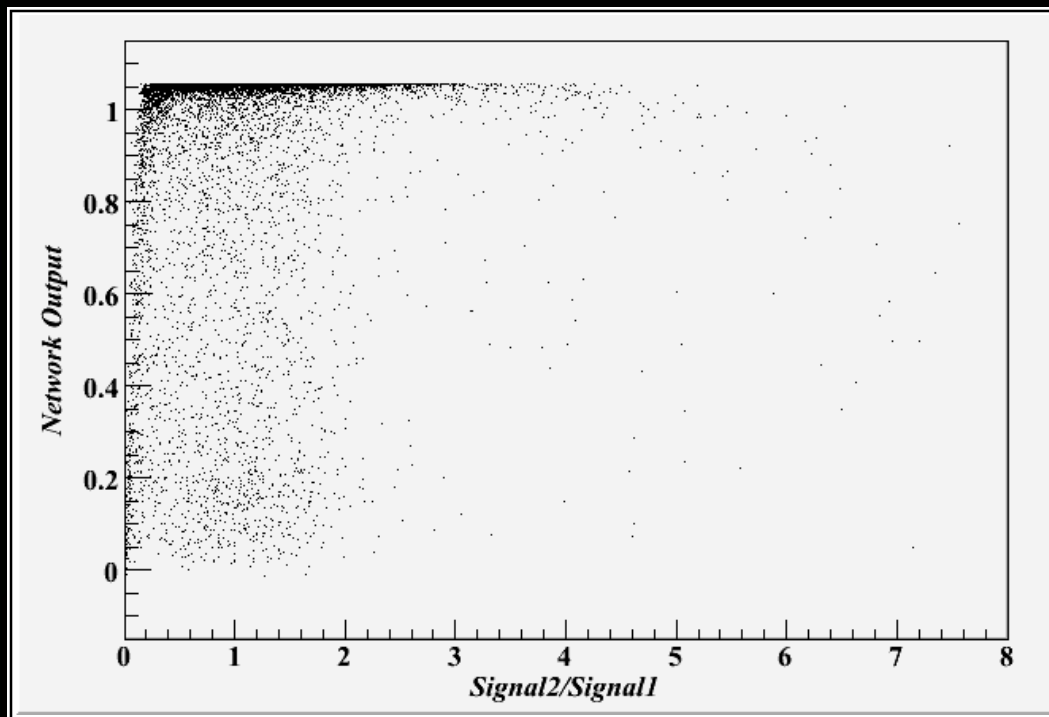
- Original training set used events over the full range of the detector
- MSE: 0.028027
- Correlation Coefficient: 0.929623
- MDL: -95361.128462
- Akaike Information Criterion: -1026929.041



The Double Hit Neural Network



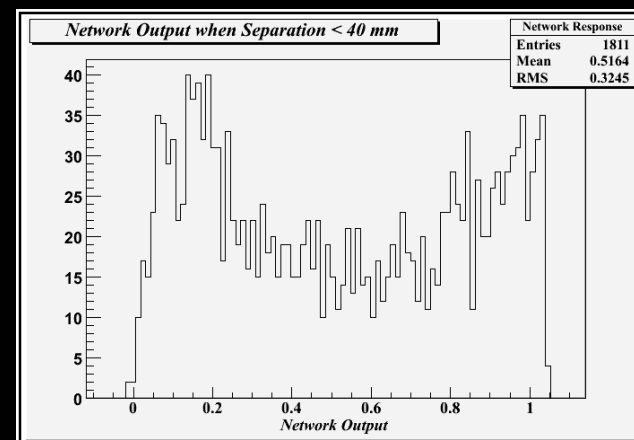
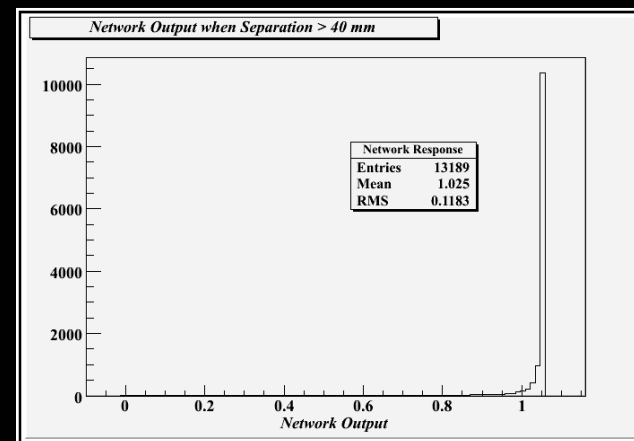
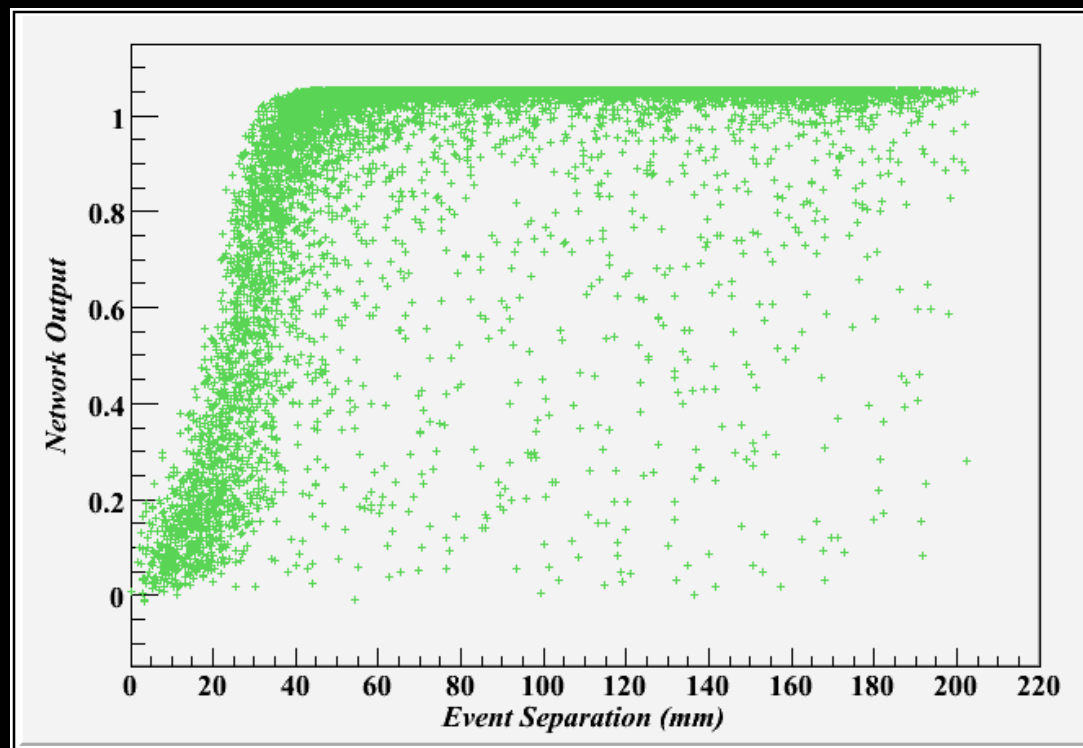
- Original training set used events over the full range of the detector



The Double Hit Neural Network



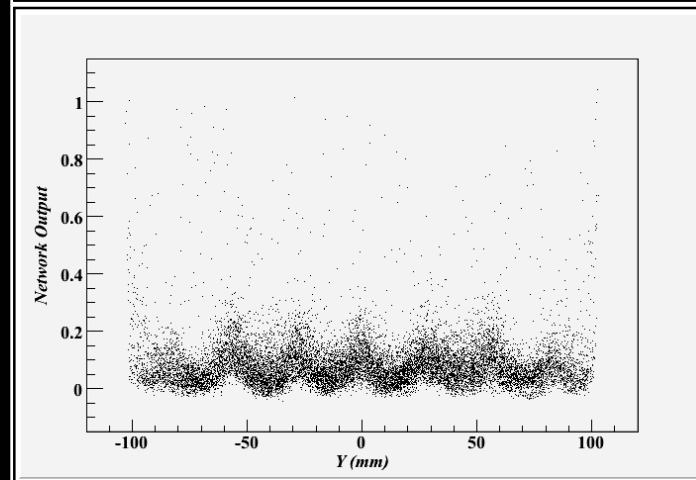
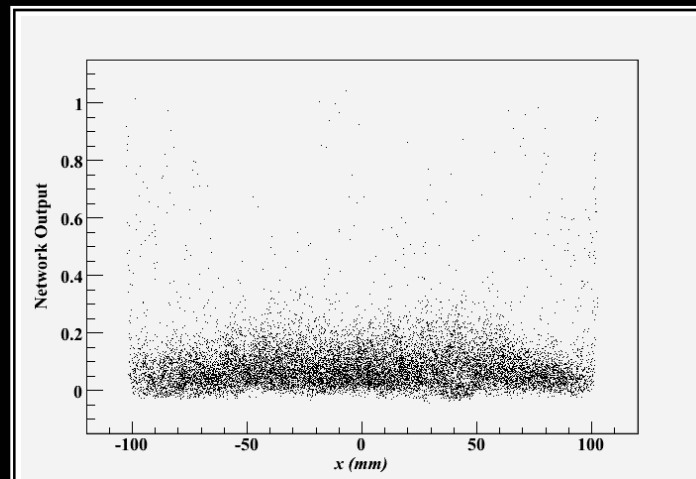
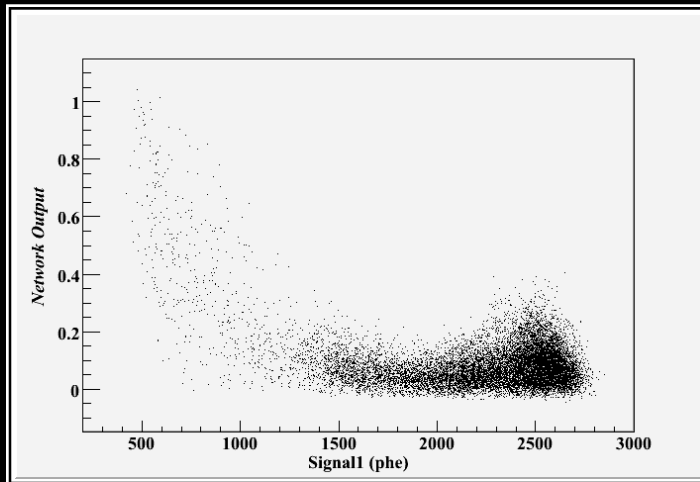
- Original training set used events over the full range of the detector



The Double Hit Neural Network



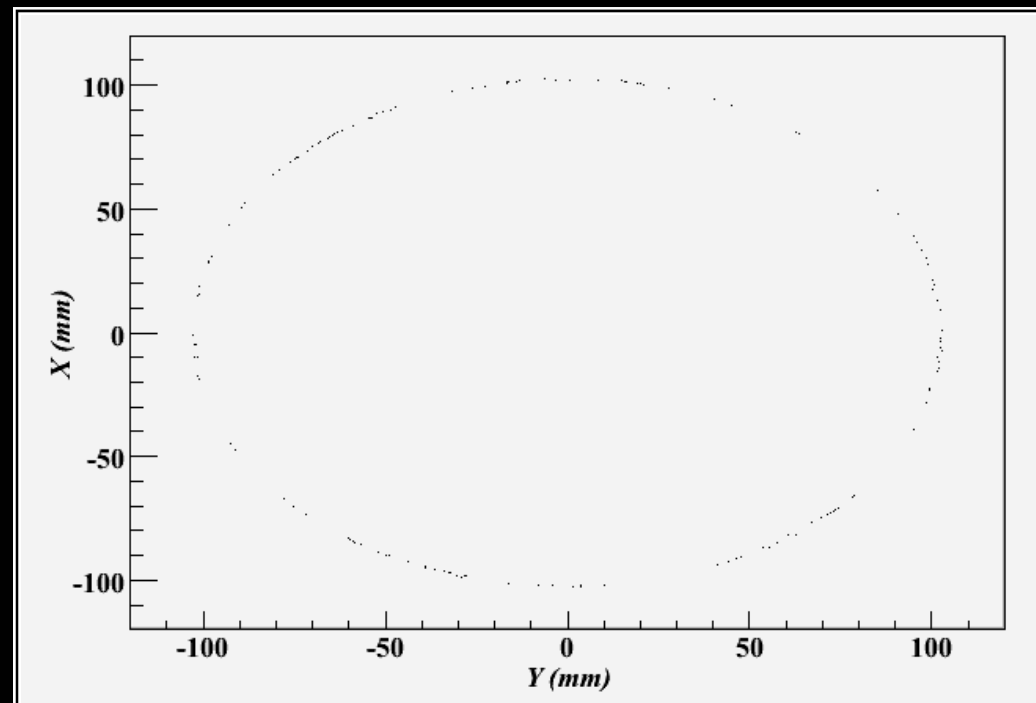
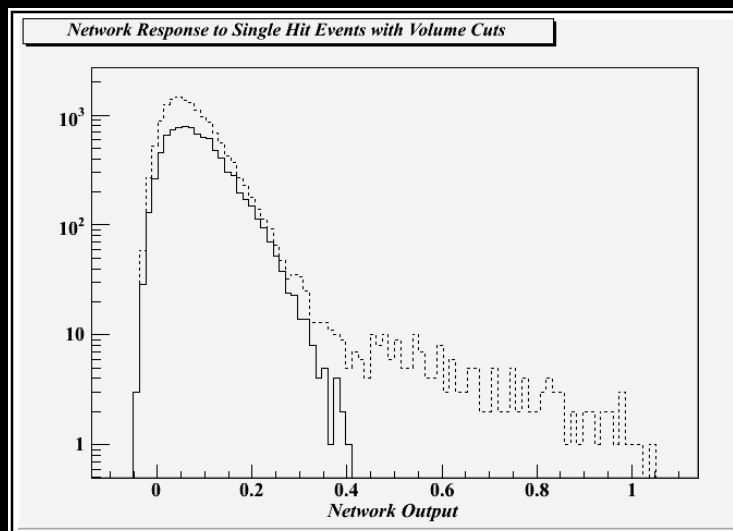
- Original training set used events over the full range of the detector



The Double Hit Neural Network



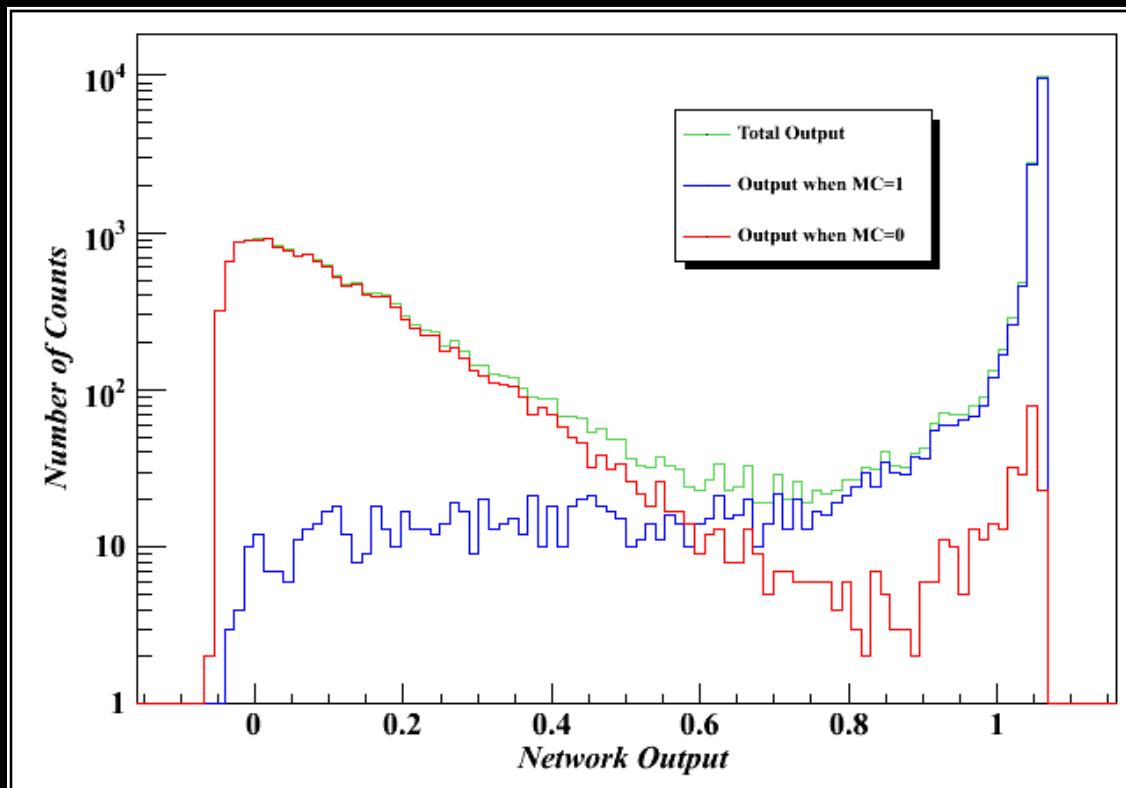
- Original training set used events over the full range of the detector



The Double Hit Neural Network



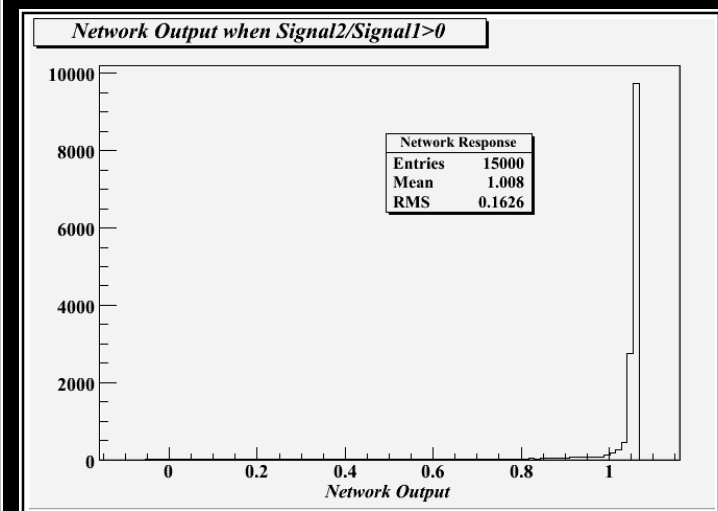
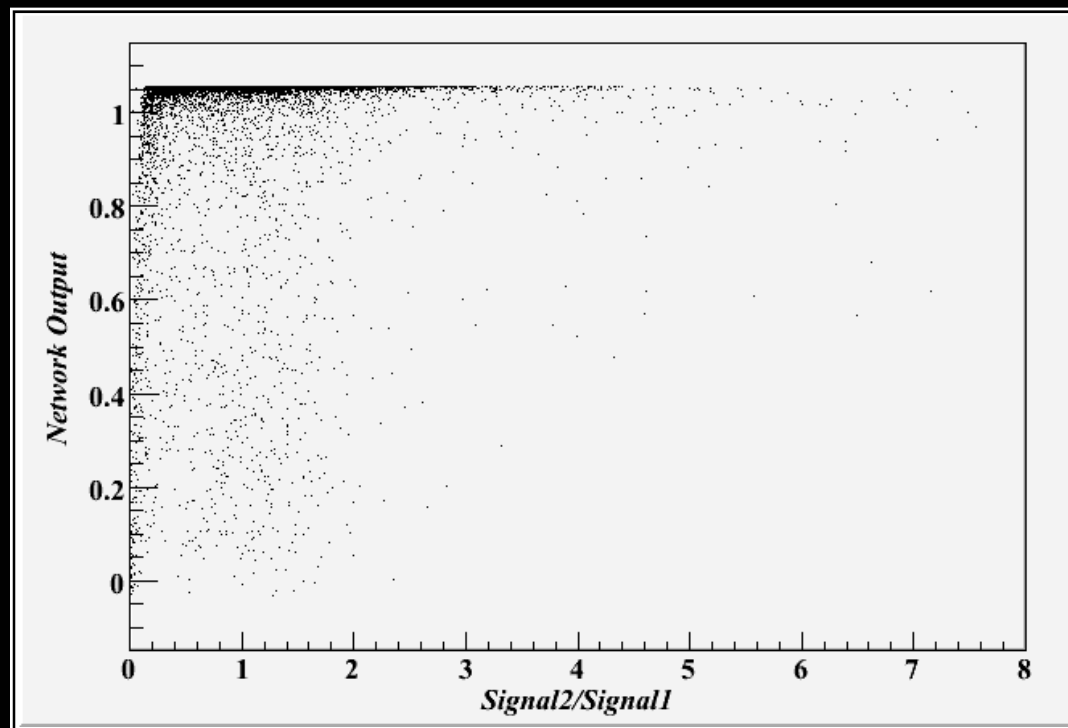
- Updated training set only used events separated by 5-40 mm
- MSE: 0.029985
- Akaike Information Criterion: -100603.3557
- Correlation Coefficient: 0.93275
- MDL: -93335.44232



The Double Hit Neural Network



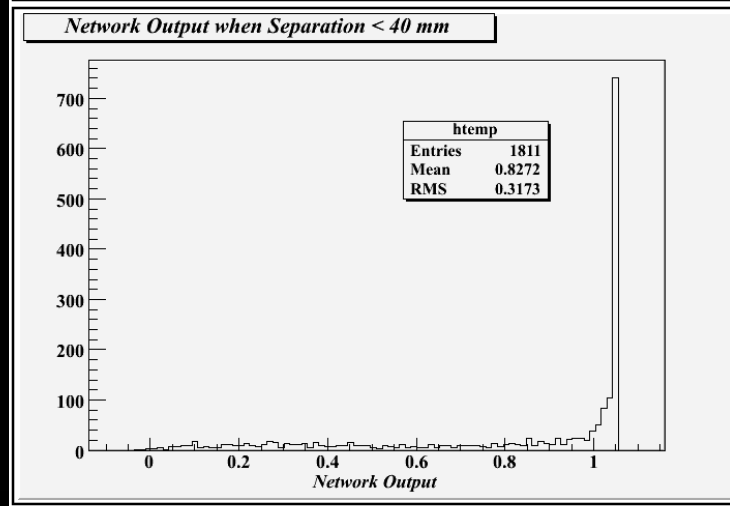
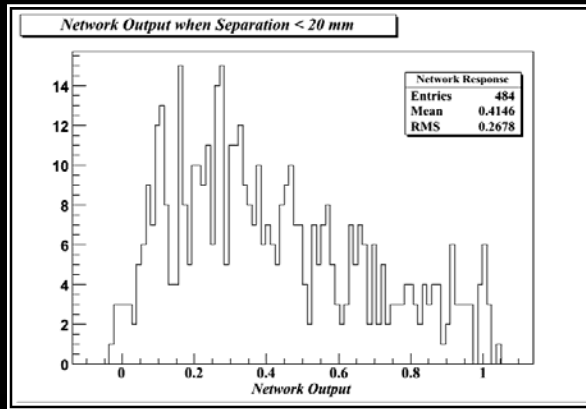
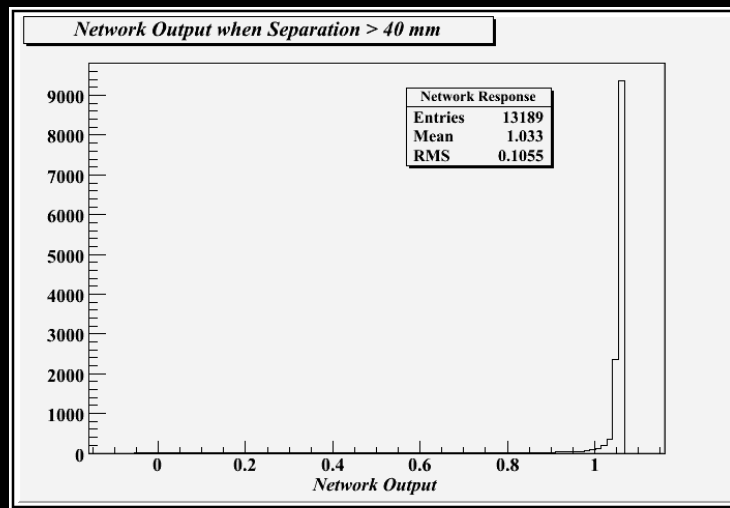
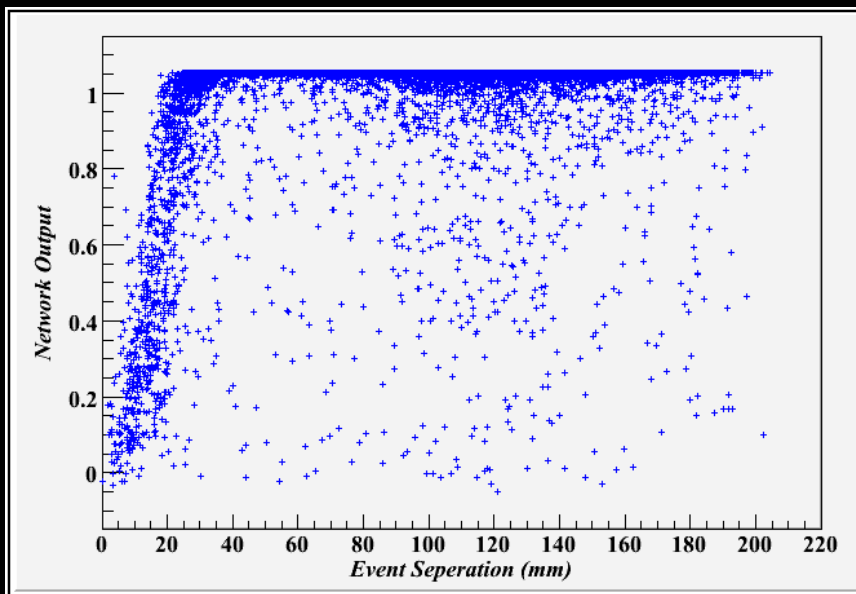
- Updated training set only used events separated by 5-40 mm



The Double Hit Neural Network



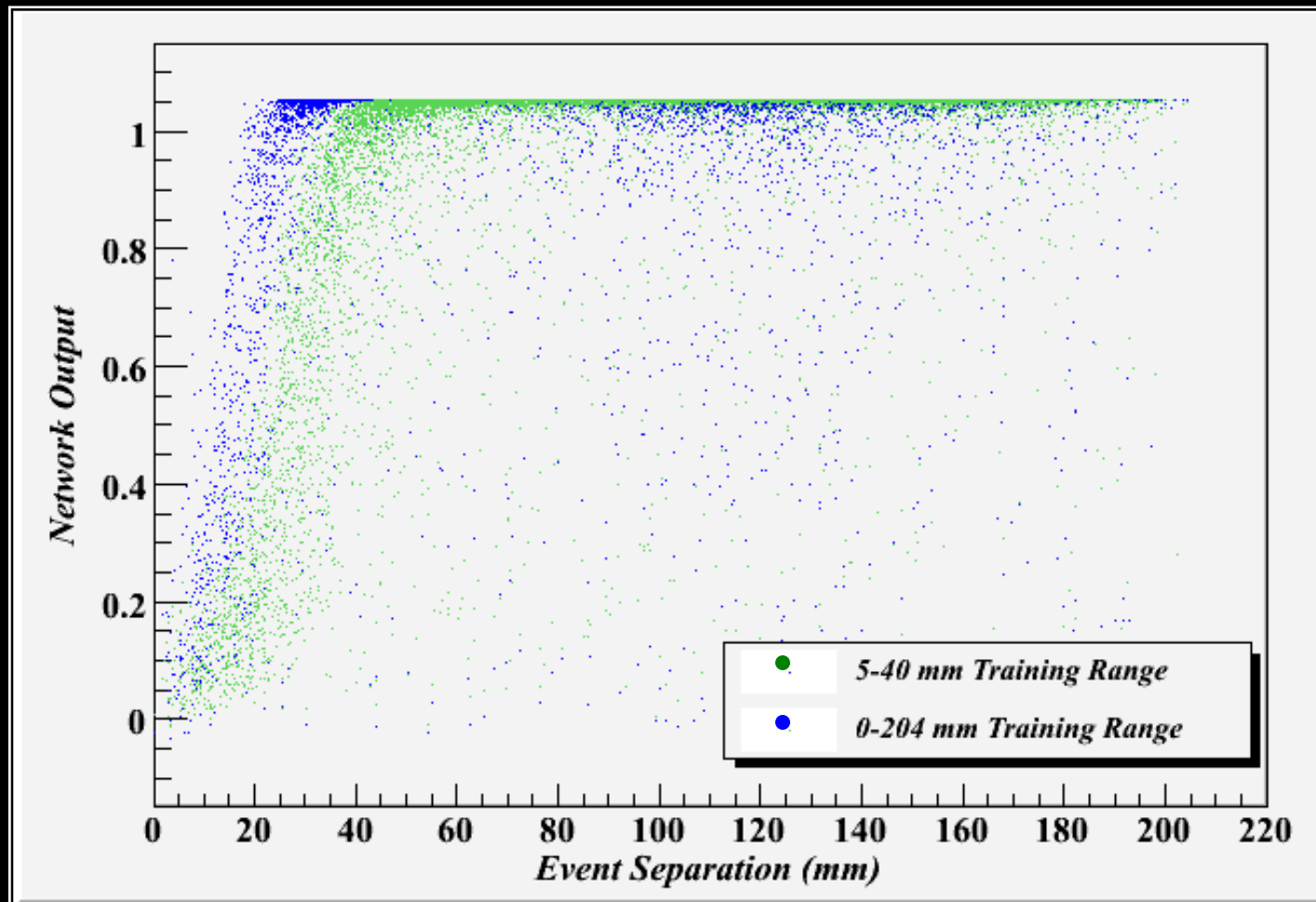
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The Double Hit Neural Network



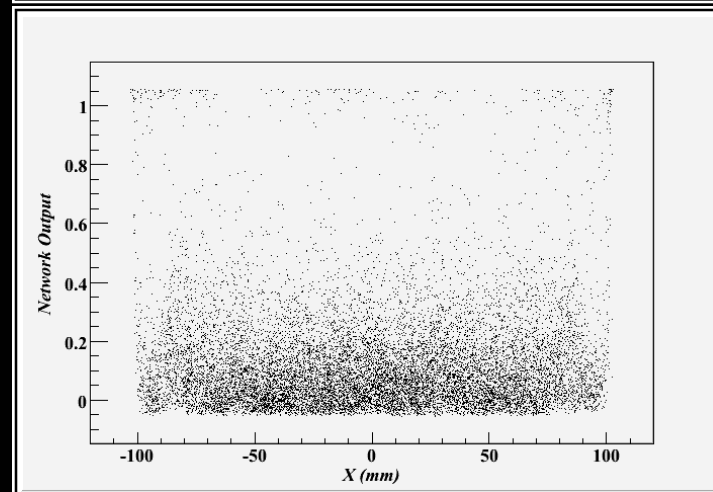
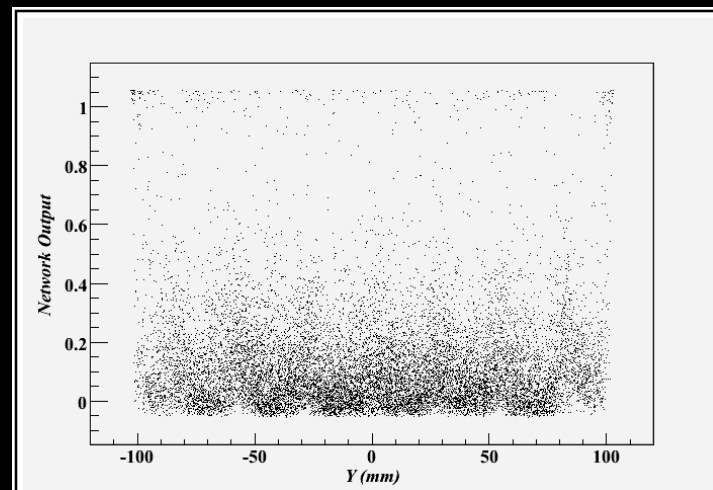
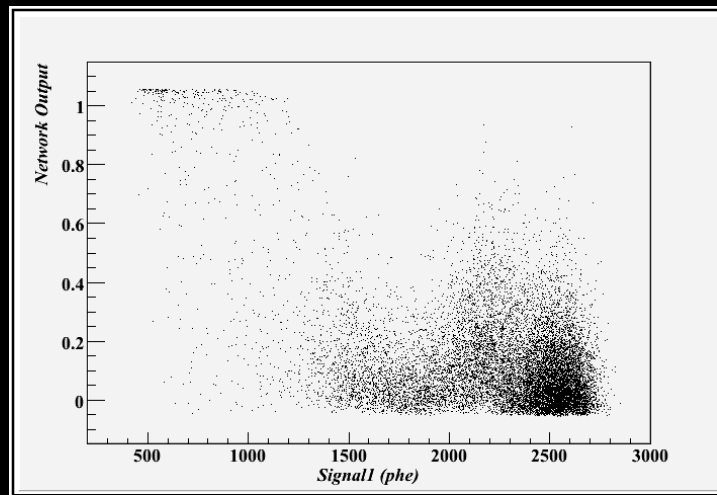
- Updated training set only used events separated by 5-40 mm



The Double Hit Neural Network



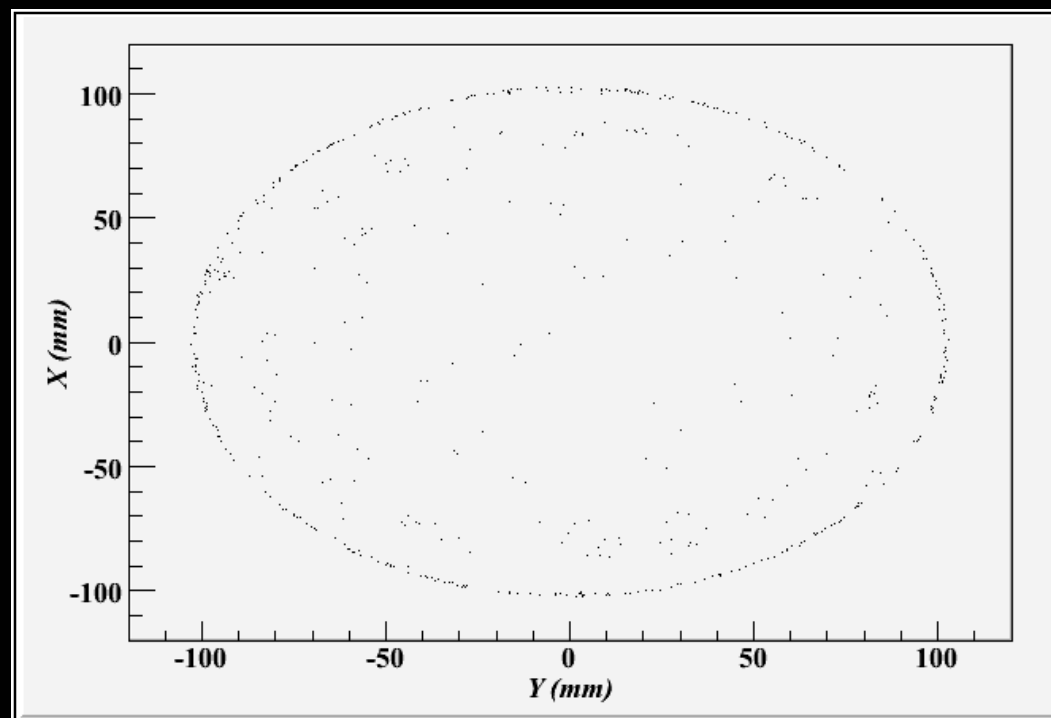
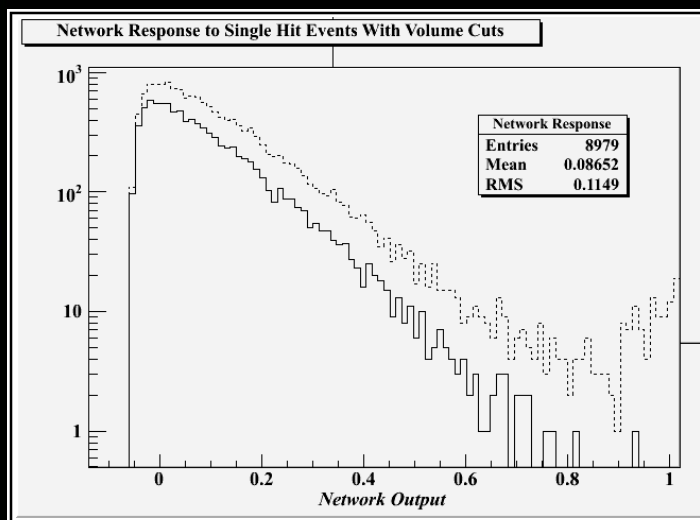
- Updated training set only used events separated by 5-40 mm



The Double Hit Neural Network



- Updated training set only used events separated by 5-40 mm

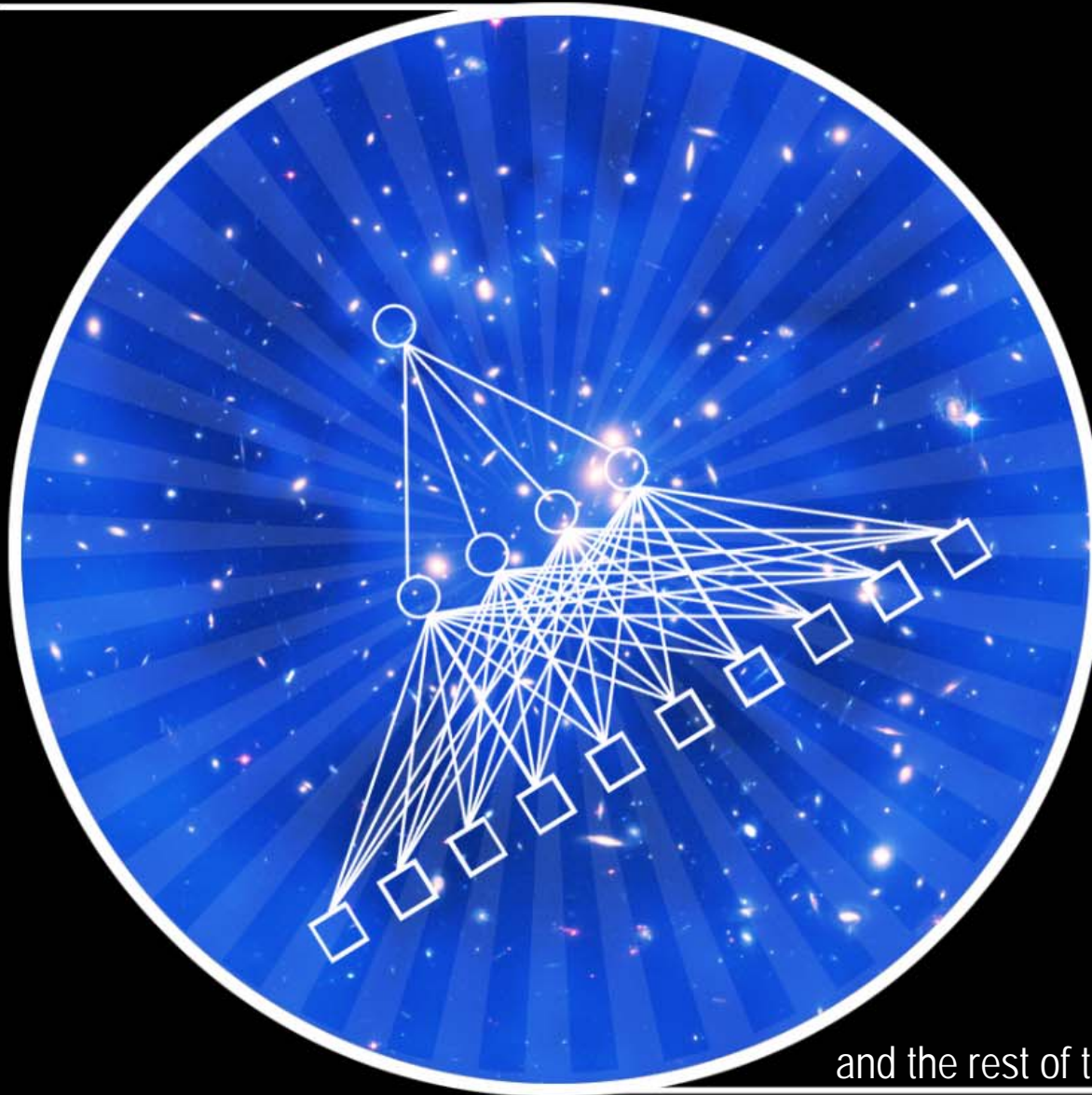


Conclusions



- Evidently the each training set provides statistically equivalent fits
 - The network trained on a smaller dataset is better at reconstructing closer events
 - The network trained on the full set less frequently interprets single hit events as double hits.
- One or both can be employed to improve background reduction
- Future of the Double Hit Network
 - Preprocessing script to change the data into a linearly separable problem or reduce the necessary number of neurons
 - Bright-space
 - Grouping PMT's
 - Reconstructing event locations of double hits

THE END



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THANK YOU