

Group 15 - LAB 1

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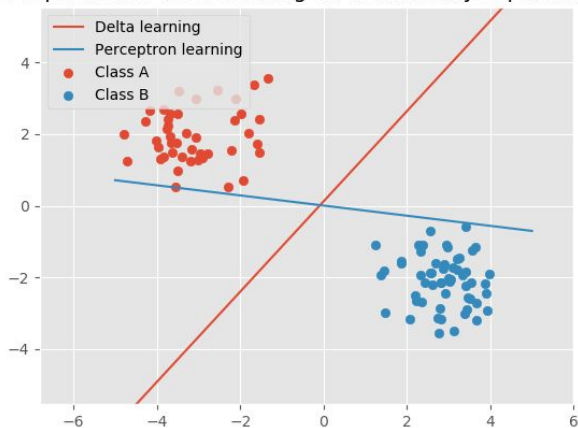


3.1

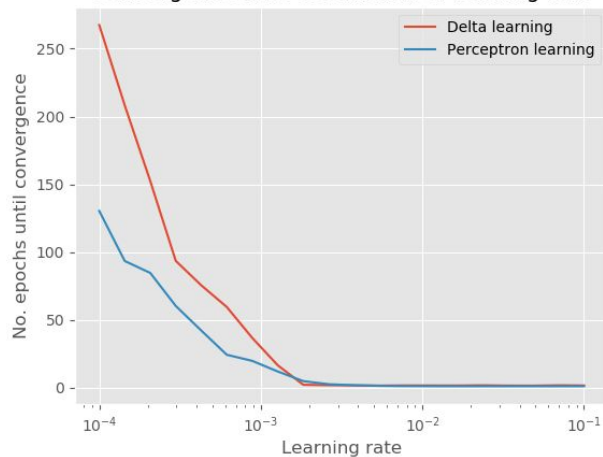
Single Layer Perceptron

Perceptron learning vs Delta rule

Perceptron and delta learning on non-linearly separable set



Convergence time as function of learning rate

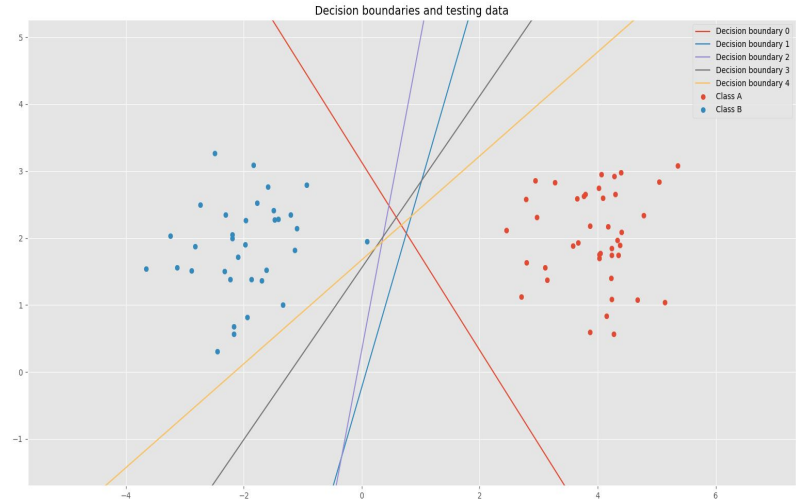
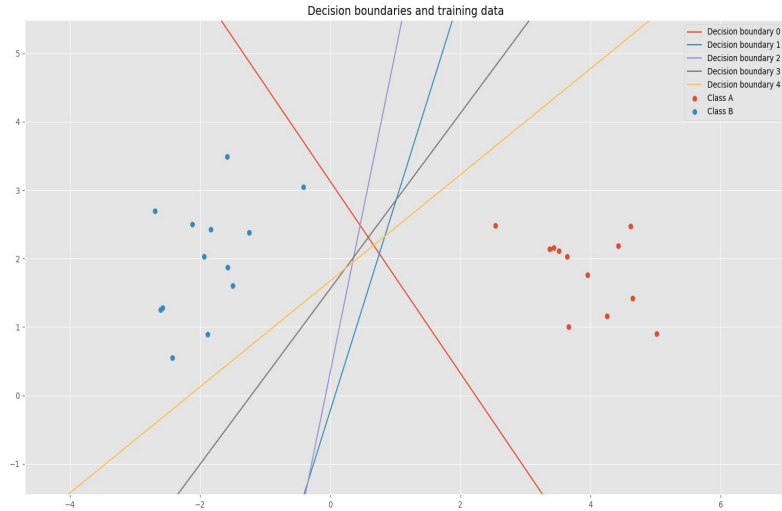


Comparison between delta learning and perceptron learning on a single dataset.



Delta rule

Sequential Learning

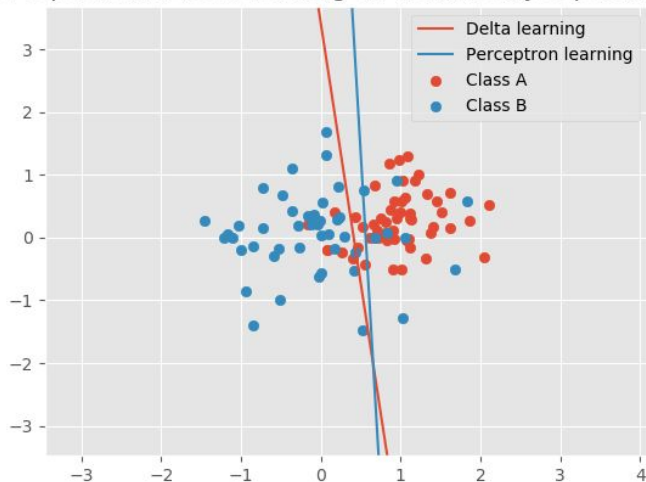


Delta learning rule applied to a single layer network using sequential learning rule instead of batch learning. (Epochs = 100 and learning rate = 0.001).

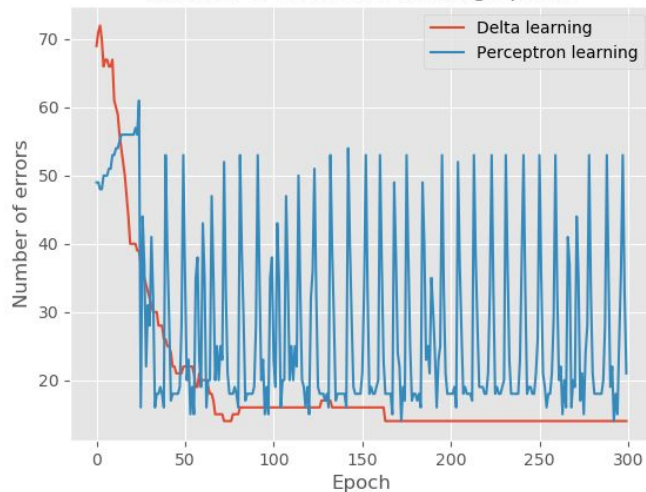
Conclusions: The network performed well in terms of classifying both the train data as well as the test data with an accuracy of 100%.

Non linearly separable

Perceptron and delta learning on non-linearly separable set



Number of errors vs. training epochs



Comparison between perceptron learning and delta learning on linearly non-separable dataset.



Subsampling

Part of A	Part of B	Mean accuracy* \pm 1std.
25%	25%	0.76 \pm 0.06
50%	0%	0.58 \pm 0.09
0%	50%	0.43 \pm 0.08
{20% A[0] > 0} \cup {80% A[0] < 0}	0%	0.23 \pm 0.00

*Averaged over 100 runs.

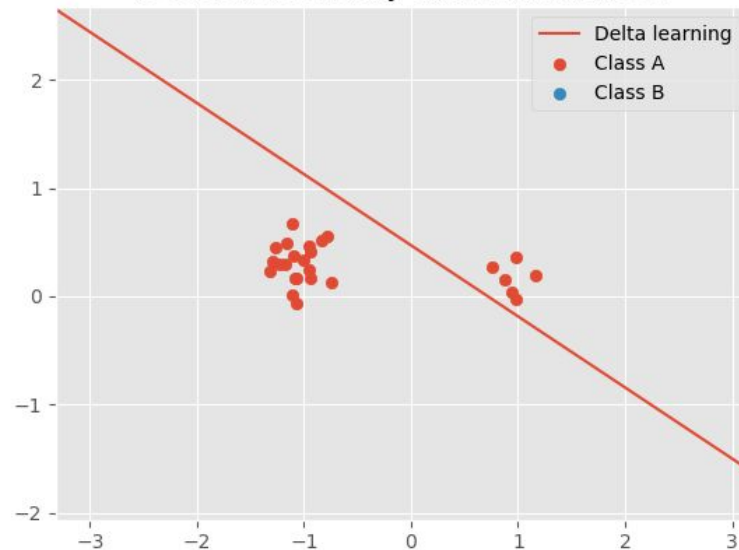


Subsampling

Decision boundary and training set



Decision boundary and validation set

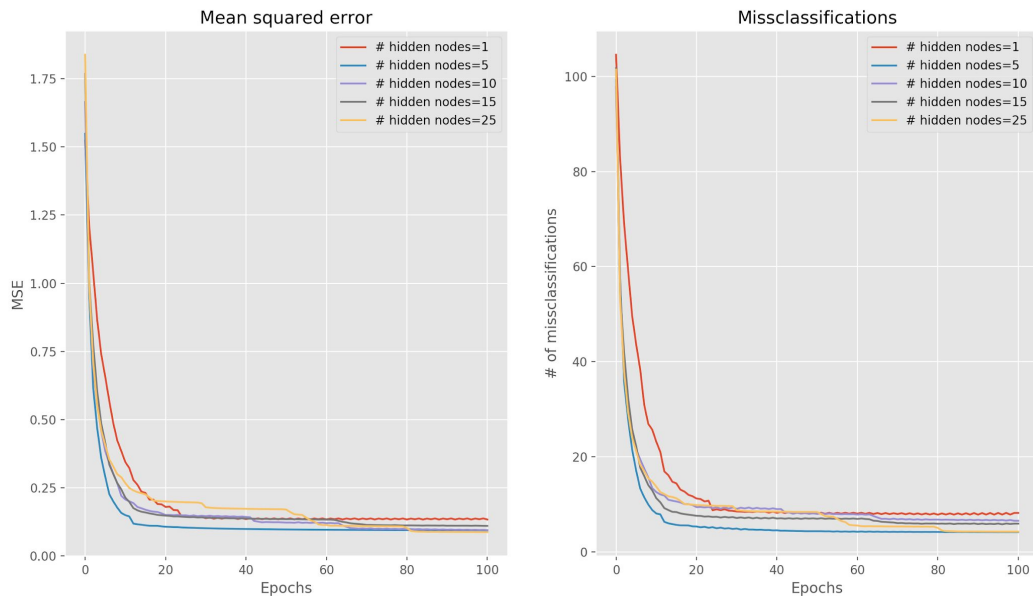


3.2

Multi layer perceptron



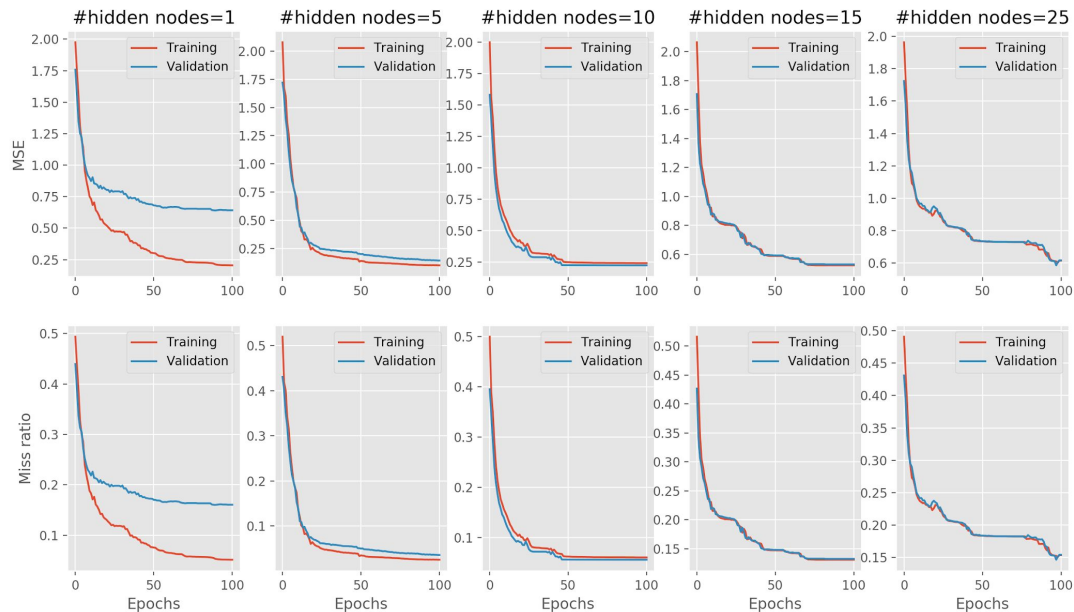
Classification of non linearly separable



Average over 100 runs for different number of nodes in the hidden layer.

Conclusions: Fastest convergence and best performance with 5 hidden nodes.

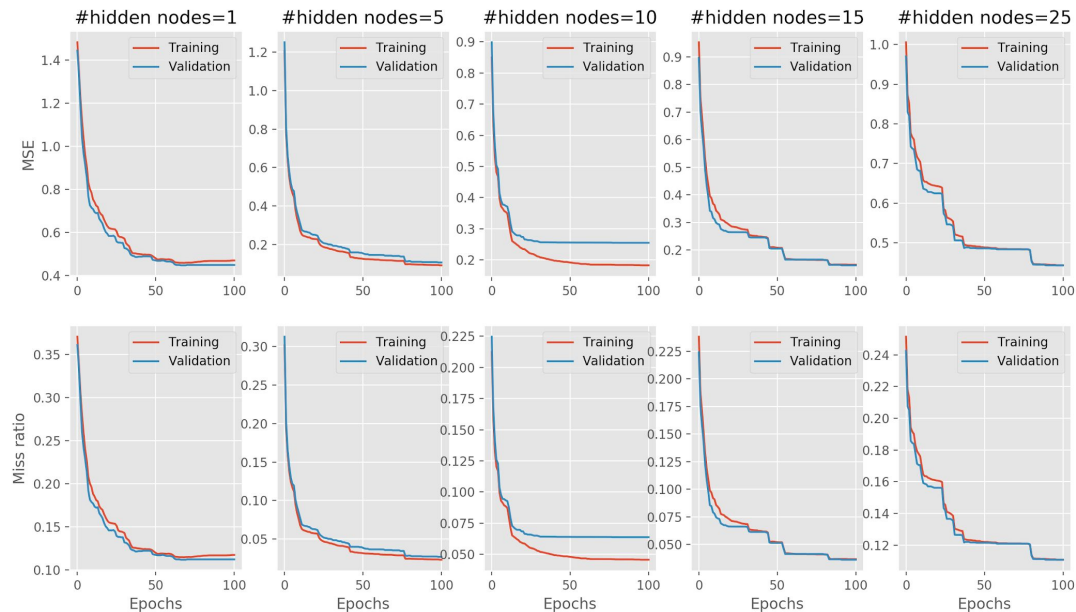
Generalisation



Average over 50 runs for different number of nodes in the hidden layer for training and validation sets.

Conclusions: Generalization is achieved for more than 10 hidden nodes.

Batch vs sequential



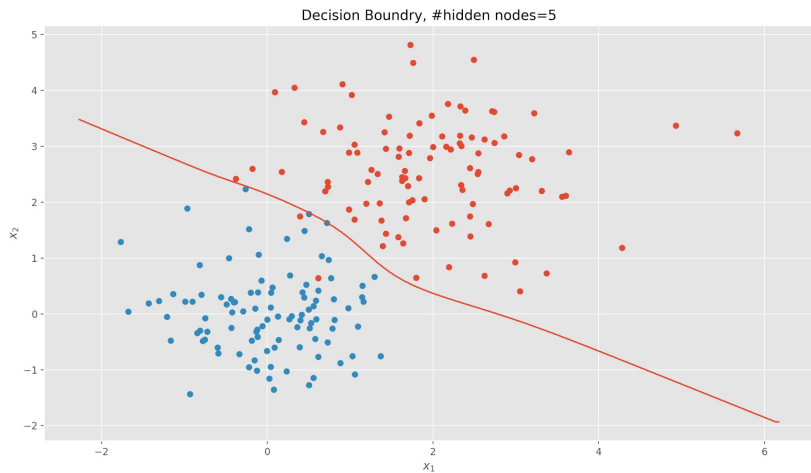
Same as previous slide, but using sequential learning.

Conclusions: Difficult to deduce differences.

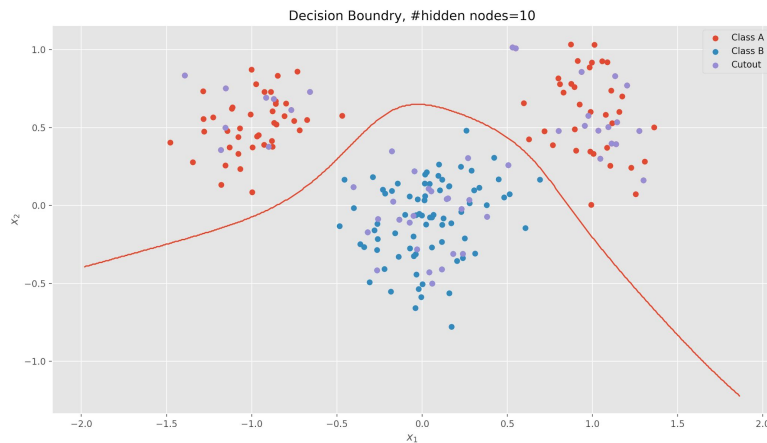


Decision boundary

With 5 hidden nodes and learning rate $\eta=0.01$ we get



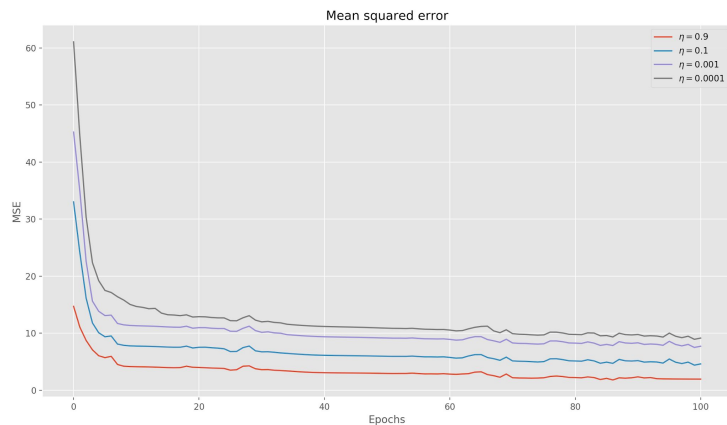
2 misclassifications with $MSE=0.05$ for the regular case



0 misclassifications with $MSE=0$ for training data and $MSE=0$ for the special case



Encoder problem 3 hidden layers



$\mathbf{X} = [-1, -1, 1, -1, -1, -1, -1, -1]$

$\mathbf{H} = [-1, -1, -1]$

$\mathbf{X} = [-1, -1, -1, -1, -1, 1, -1, -1]$

$\mathbf{H} = [1, 1, -1]$

$\mathbf{X} = [1, -1, -1, -1, -1, -1, -1, -1]$

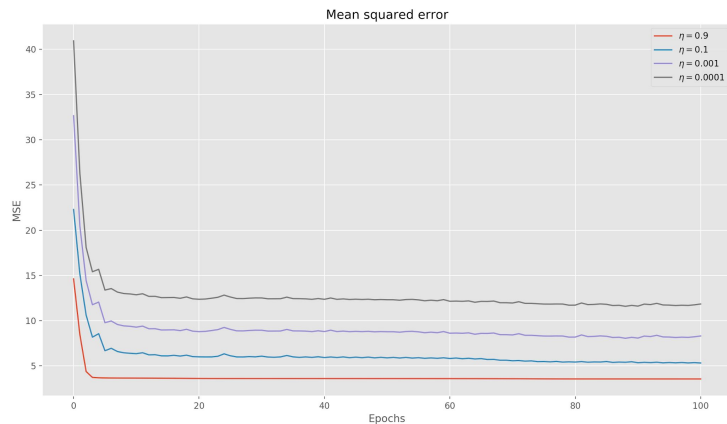
$\mathbf{H} = [1, 1, 1]$

Converges for high η .

Conclusions: Fastest convergence and best performance with 5 hidden nodes.



Encoder problem 2 hidden layers



$\mathbf{X} = [-1, -1, 1, -1, -1, -1, -1, -1]$

$\mathbf{H} = [-1, -1]$

$\mathbf{X} = [-1, -1, -1, -1, -1, 1, -1, -1]$

$\mathbf{H} = [-1, -1]$

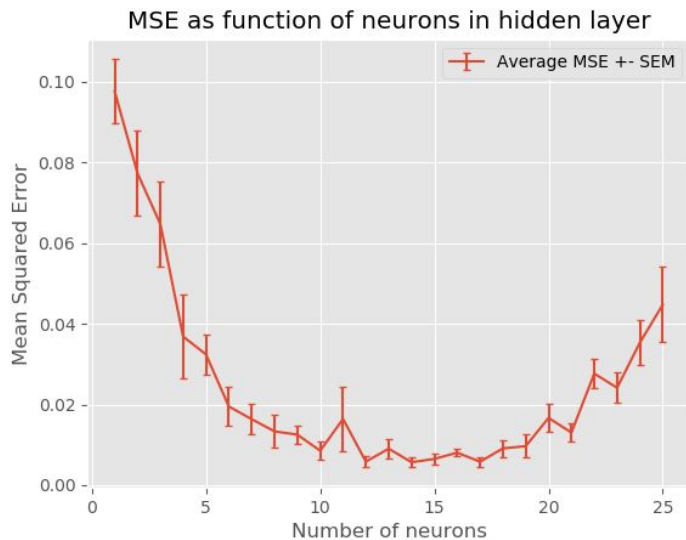
$\mathbf{X} = [1, -1, -1, -1, -1, -1, -1, -1]$

$\mathbf{H} = [1, -1]$

Converges for high η .

Conclusions: Fastest convergence and best performance with 5 hidden nodes.

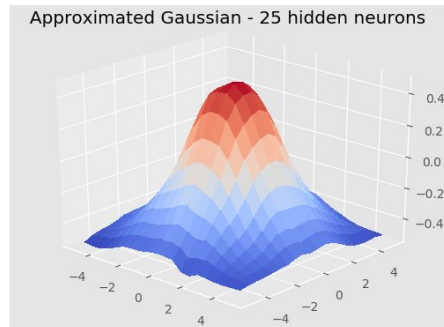
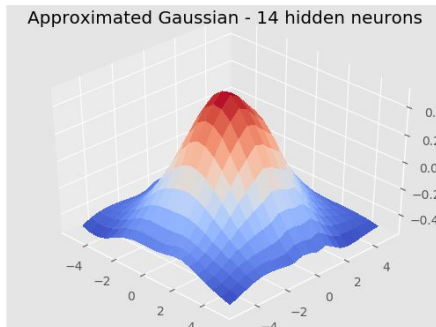
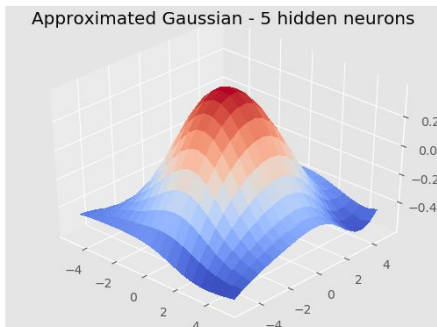
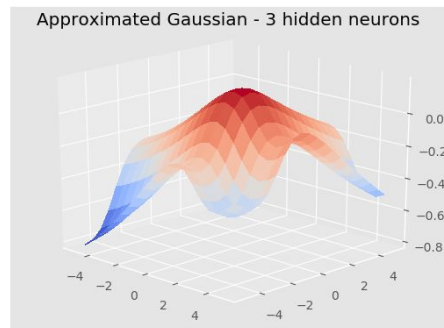
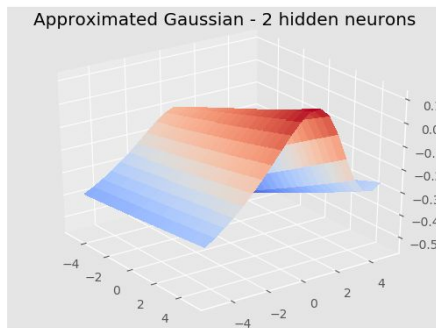
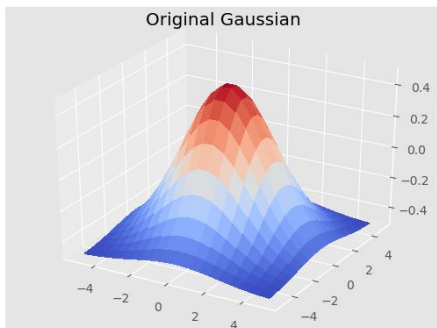
Function approximation



Left: Average MSE over 10 runs for different number of nodes in the hidden layer.

Right: Average MSE over 10 runs for different fractions of samples in training data.

Function approximation



Surface plots of the approximated gaussian, with training fraction 0.7.

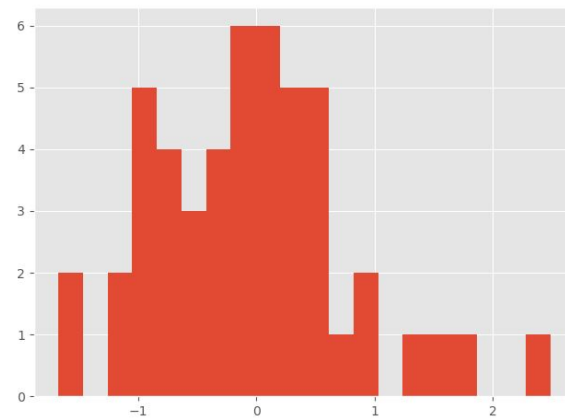
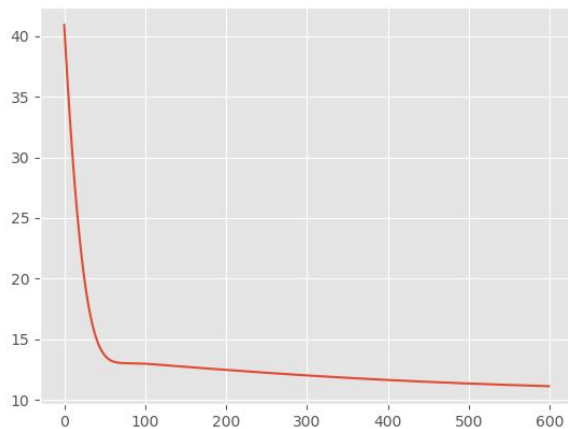
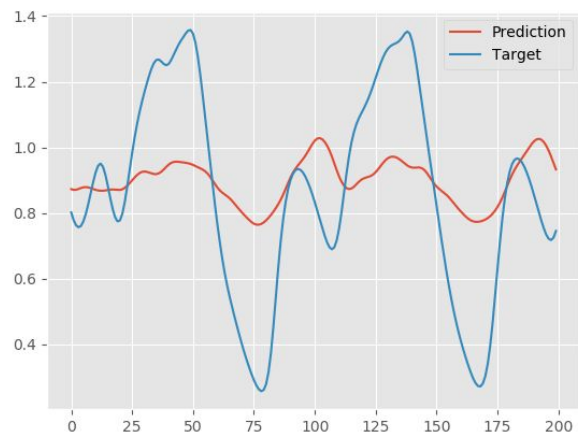


4

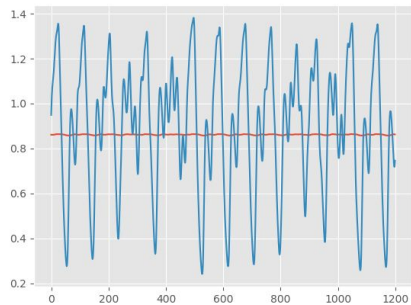
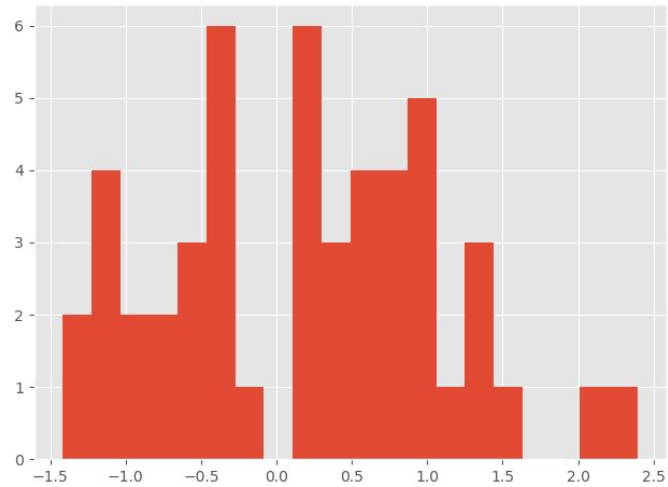
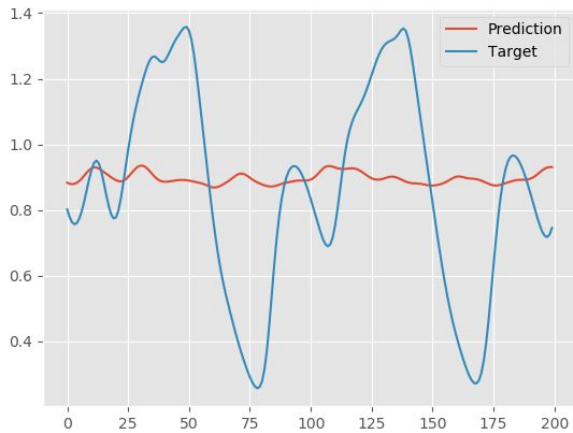
Time series prediction



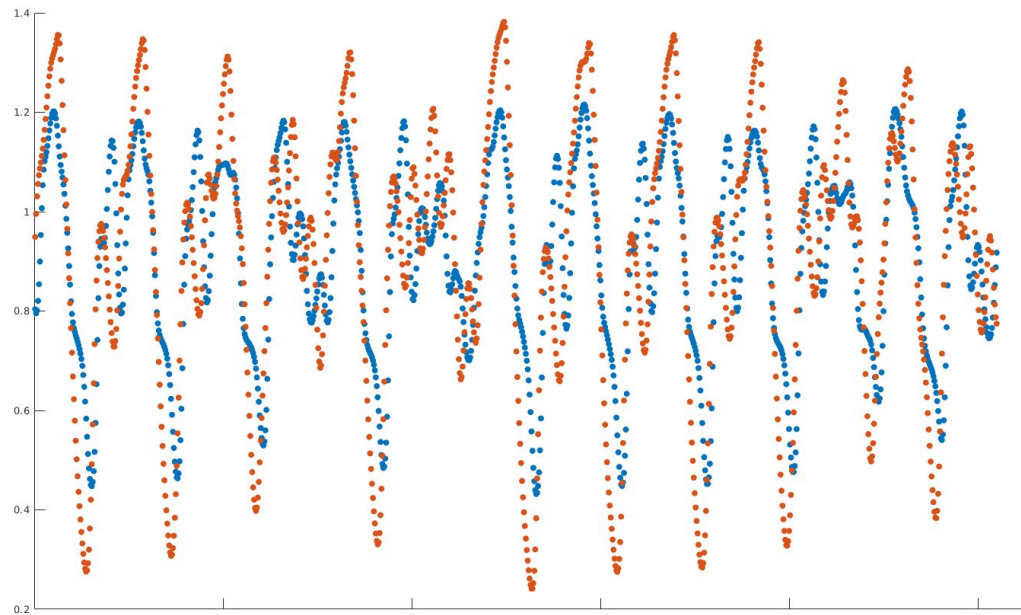
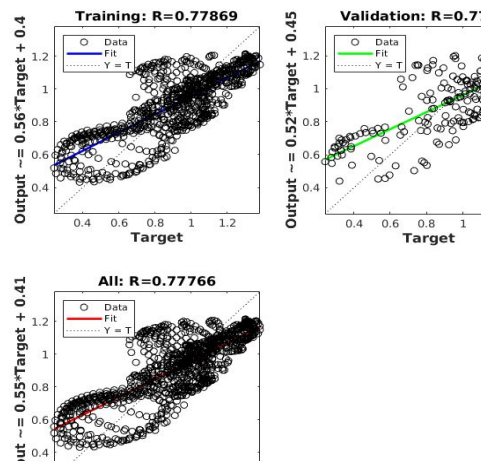
Three-layer perceptron



Three-layer perceptron

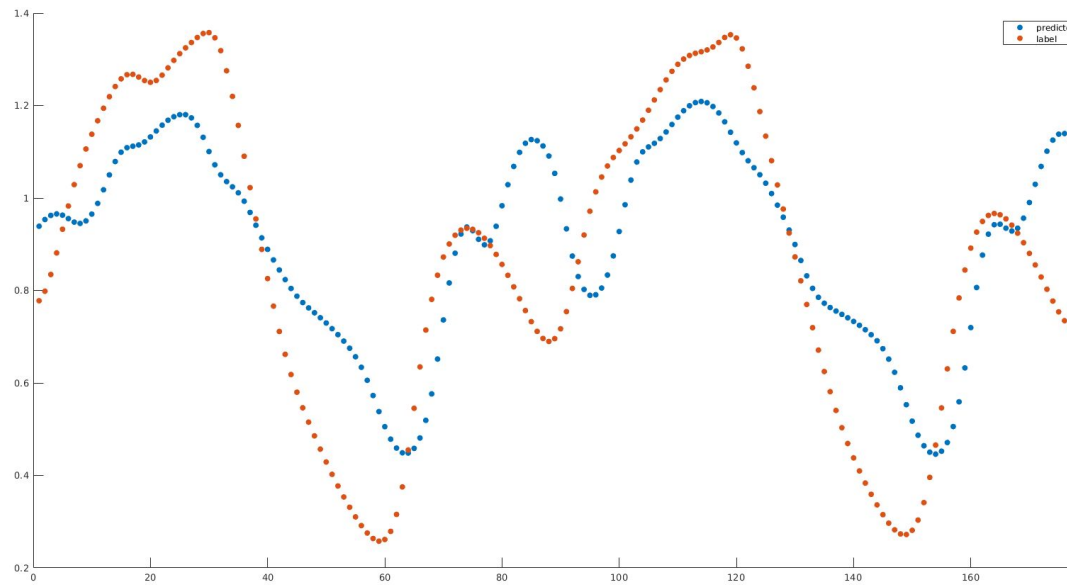
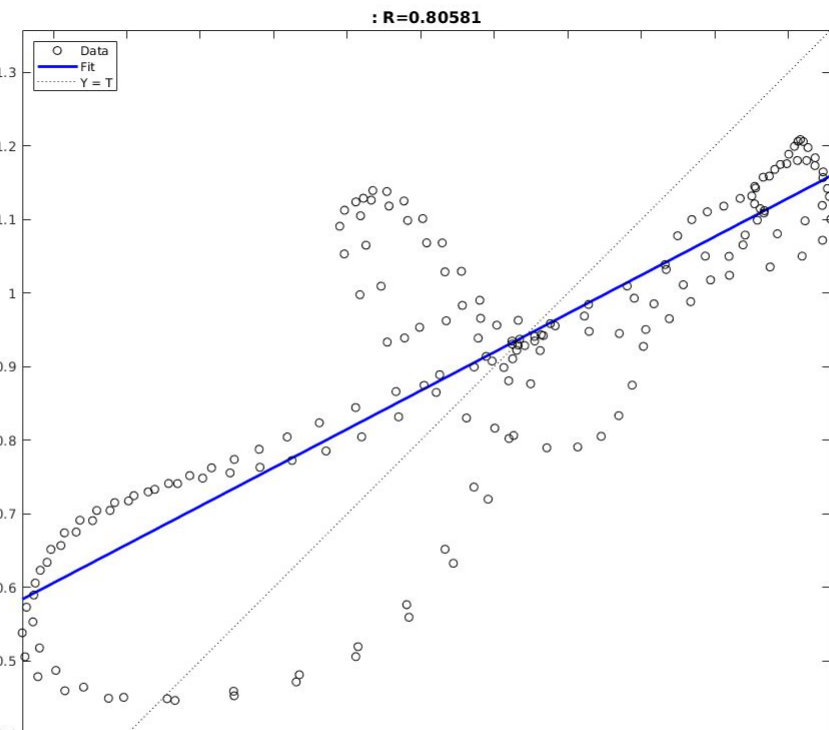


Three-layer perceptron



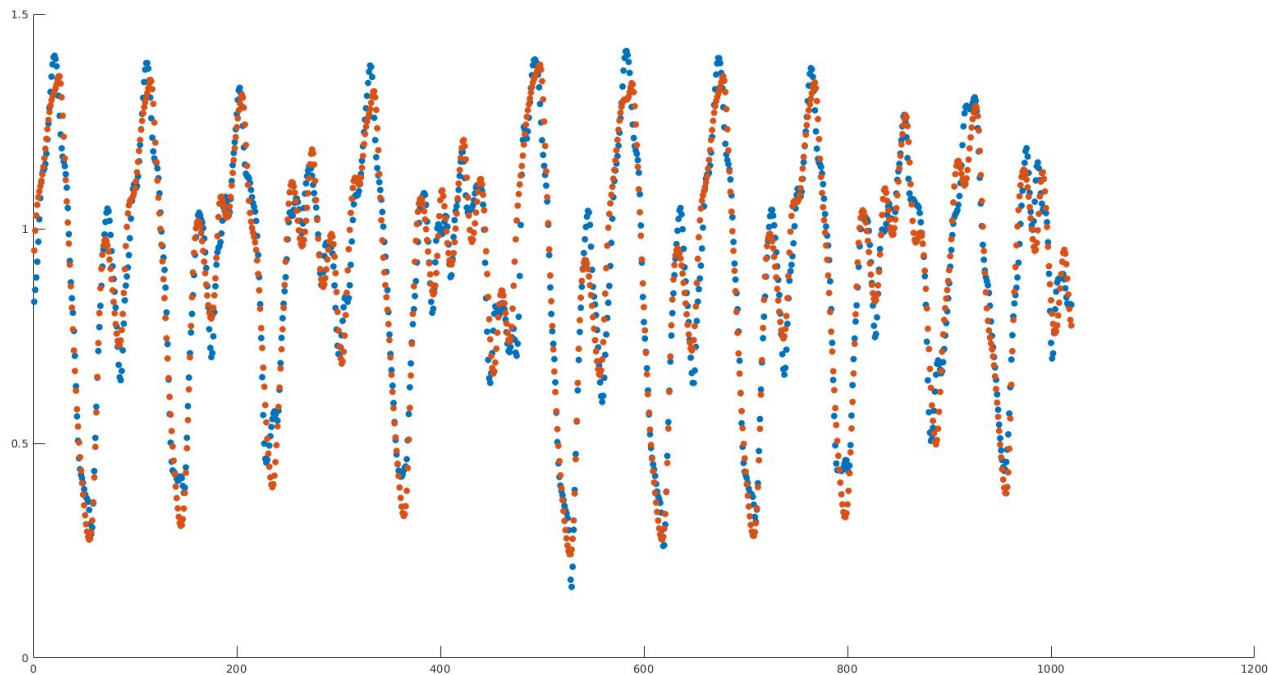


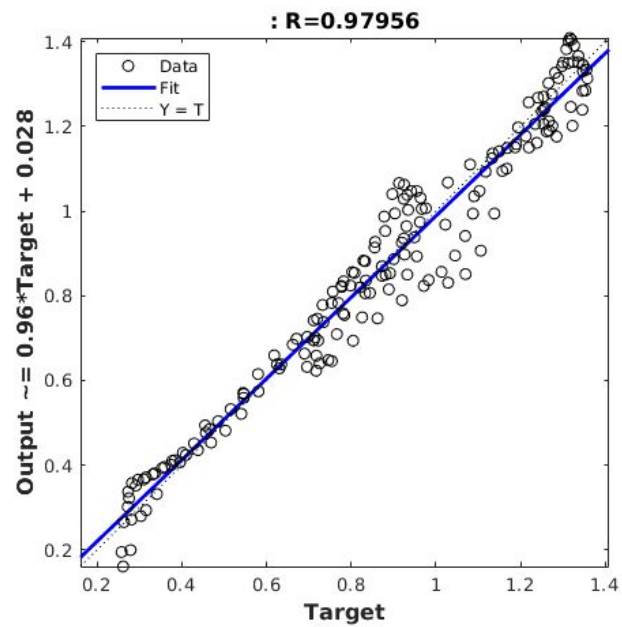
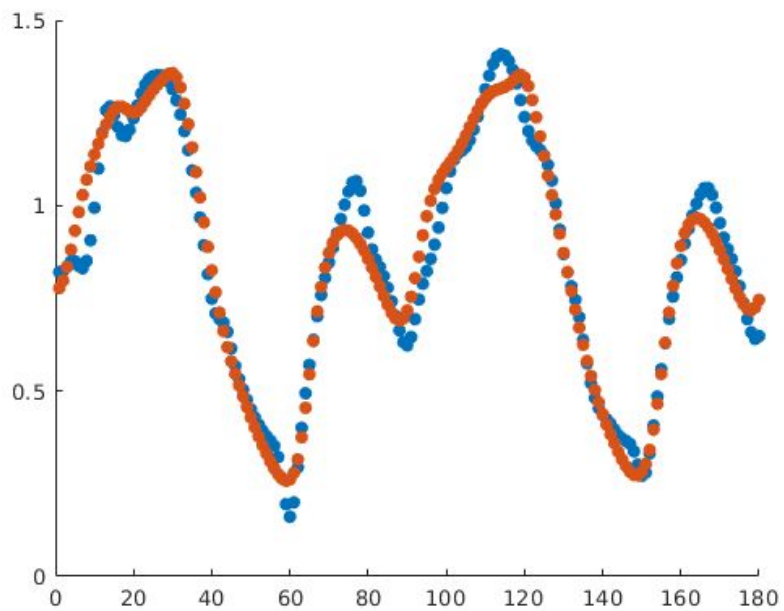
Three-layer perceptron





Three-layer perceptron









Three-layer perceptron

Average over 100 runs for different number of nodes in the hidden layer.

Conclusions: Fastest convergence and best performance with 5 hidden nodes.