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# SUPERVISED LEARNING AND NEURAL NETWORKS

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COMP3611: Coursework 1



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## 1) THE DATASET

### Class 1

x-range = [2,5]  
y-range = [1,4]  
rotation = 75 degrees  
colour = red

### Class 2

x-range = [1, 3]  
y-range = [-5, -1]  
rotation = 75 degrees  
colour = blue

### Class 3

mean = (-2, -3)  
covariance =  $\begin{bmatrix} 0.5 & 0 \\ 0 & 3 \end{bmatrix}$   
colour = green

### Class 4

mean = (-4, -1)  
covariance =  $\begin{bmatrix} 3 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$   
colour = cyan

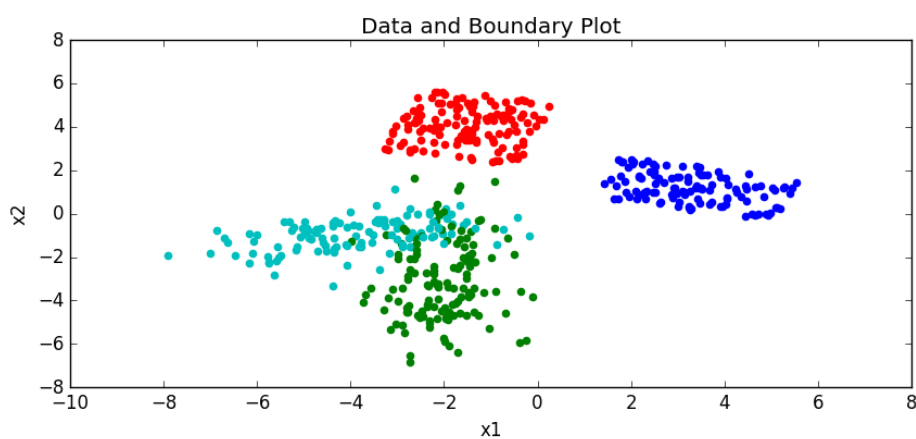


Figure 1. Plot of the data instances.

## 2) THE SINGLE-LAYER PERCEPTRON

**Classes Chosen:** Class 2 (blue) and all other classes treated as a single class (red, green, cyan).

**Reason:** The two classes as shown above are linearly separable hence the single perceptron will be able to linearly distinguish them.

On the other hand, if classes, say, Class 3 (cyan) and Class 4 (green) were chosen, then the single-layer perceptron would perform poorly as they are not linearly separable.

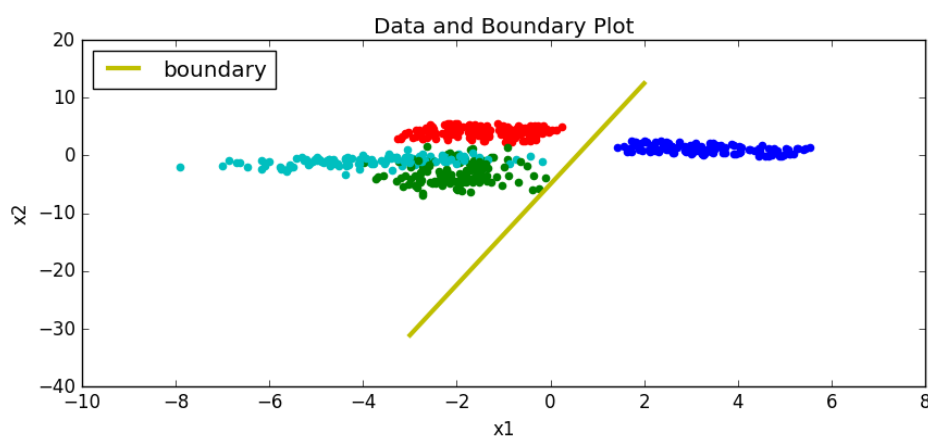
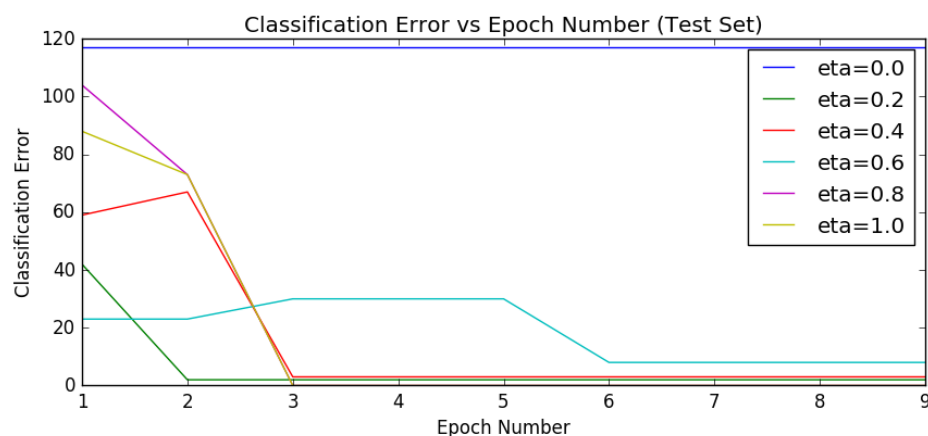
### Training Parameters

epoch = 10

batch = 10

eta (optimal) = 0.2 (since it reaches near 0 error the first and has the least errors after the epoch=1)

accuracy = 1.0 = 100% (see the hyperplane below separator below correctly classifying all instances)



## 2) THE MULTI-LAYER PERCEPTRON

### Training Parameters

epoch = 200

batch = 5

eta = 0.1

hidden nodes = 2

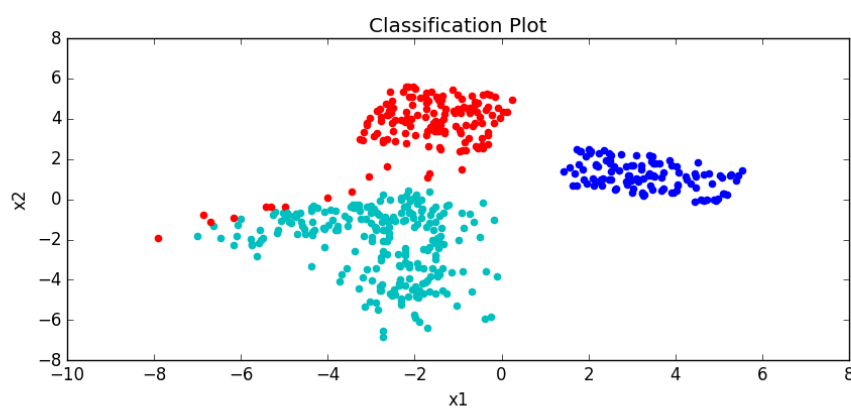
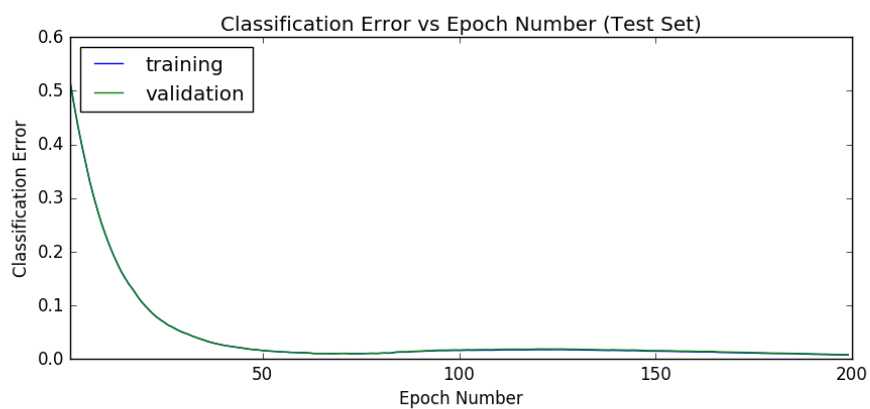
confusion matrix:

```
[[132  0  0  0]
```

```
 [ 0 116  0  0]
```

```
 [ 4  0  0 129]
```

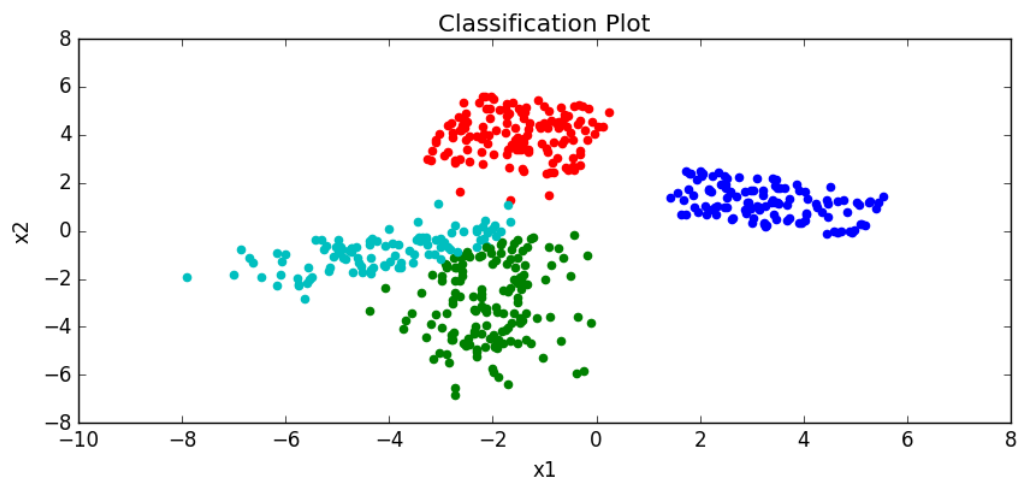
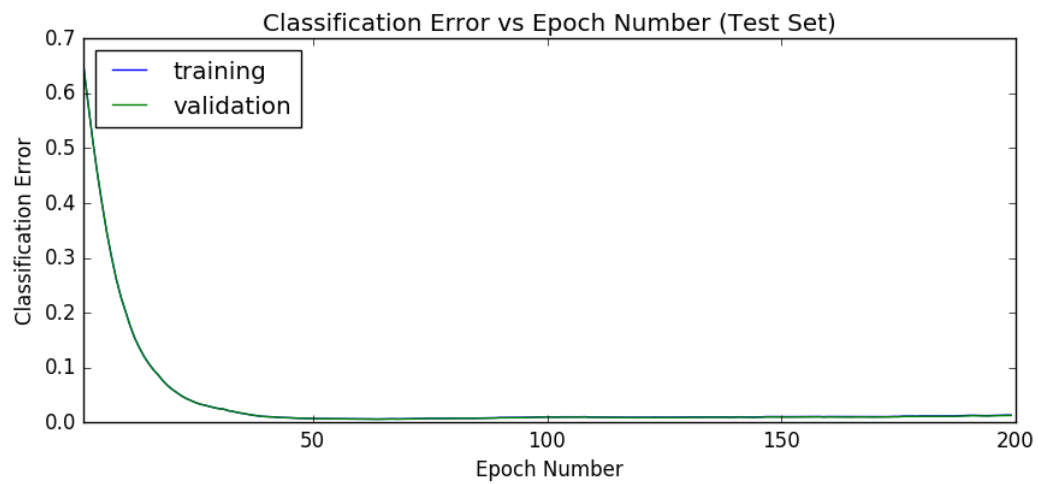
```
 [10  0  0 109]]
```



hidden nodes = 5 (**BEST PERFORMANCE (see matrix and plot below)**)

confusion matrix:

```
[[132  0  0  0]
 [  0 116  0  0]
 [  3  0 120 10]
 [  0  0  26 93]]
```



hidden nodes = 20

confusion matrix:

```
[[132  0  0  0]
 [  0 116  0  0]
 [  4  0 119 10]
 [  2  0 24  93]]
```

