

# PERCEPTRON LEARNING ALGORITHM

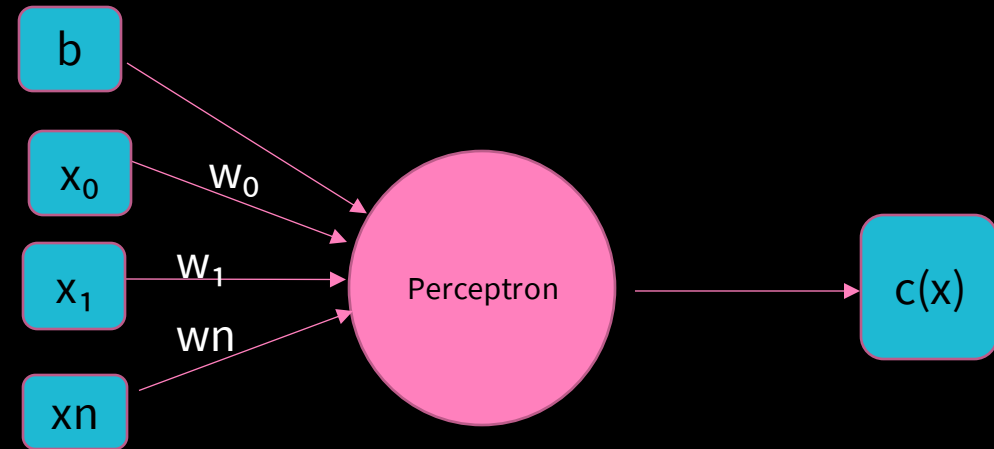
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# Perceptron Summary

- The idea of the Perceptron:

- Inputs
- Weights
- Perceptron
- Output



- $c(x) = b + x_0 * w_0 + x_1 * w_1 + \dots + x_n * w_n$





## Weight Updates

- Calculate the hypothesis  $h(x)$  of the example
  - Pos or Neg?
- If  $h(x)$  is the correct class, weights don't change
- If  $h(x)$  is greater than the class, reduce weights
- If  $h(x)$  is less than the class, increase weights
- Weights are scaled by a learning rate ( $\eta$ )





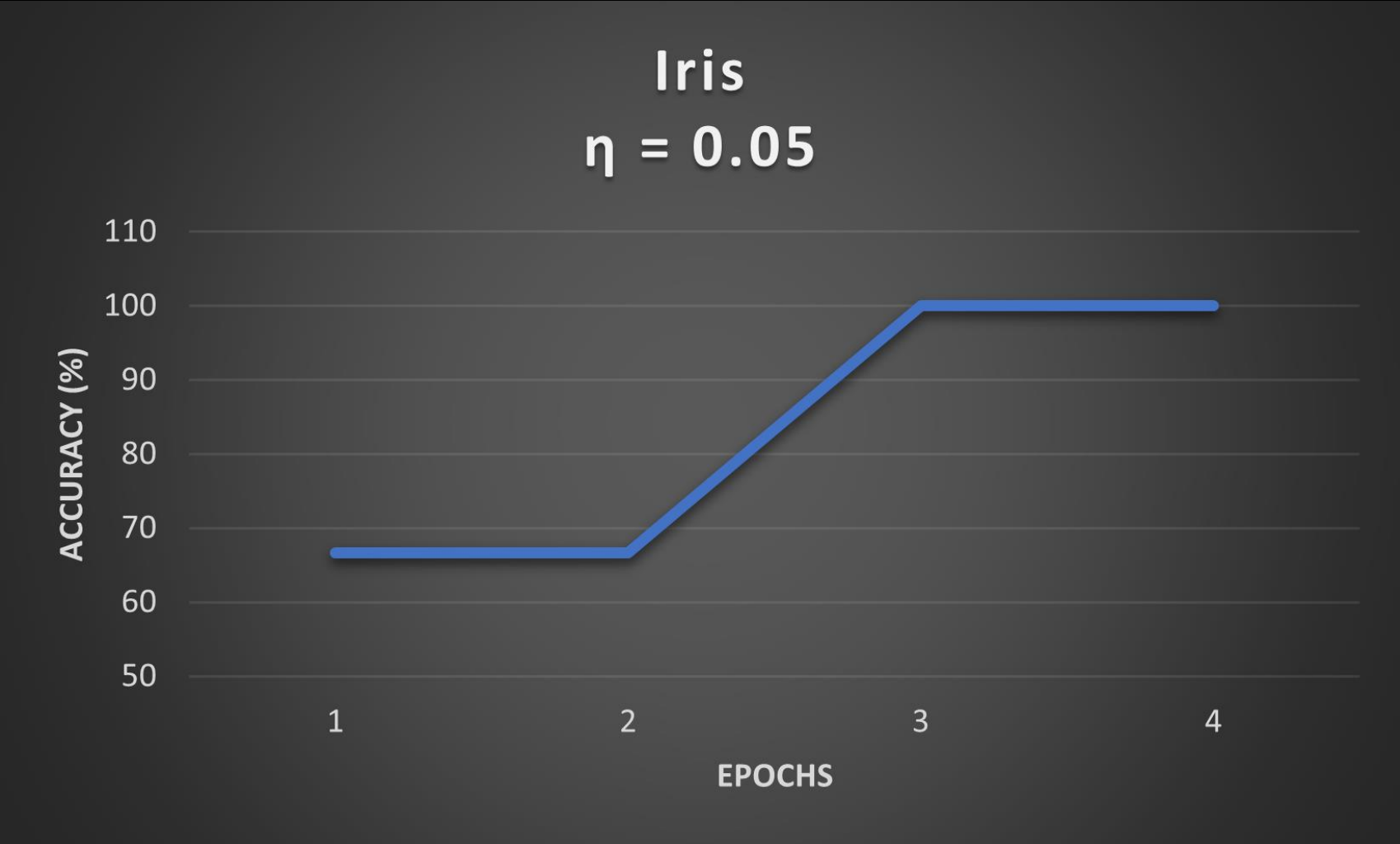
# Databases

- Iris dataset (150 examples)
  - 50 pos, 100 neg
- Balloons dataset (20 examples)
  - 8 pos, 12 neg
- Banknote dataset (1372 examples)
  - 610 pos, 762 neg





# Iris Database



# Iris Database

```
Epoch: 0  
Weights[0] = -1.65  
Weights[1] = 0  
Weights[2] = -3.95  
Weights[3] = -1.75
```

```
Epoch: 1  
Weights[0] = 0.9  
Weights[1] = 1.75  
Weights[2] = -3.25  
Weights[3] = -1.65
```

```
Epoch: 2  
Weights[0] = 0.9  
Weights[1] = 1.75  
Weights[2] = -3.25  
Weights[3] = -1.65
```

```
Epoch: 3  
Weights[0] = 0.9  
Weights[1] = 1.75  
Weights[2] = -3.25  
Weights[3] = -1.65
```

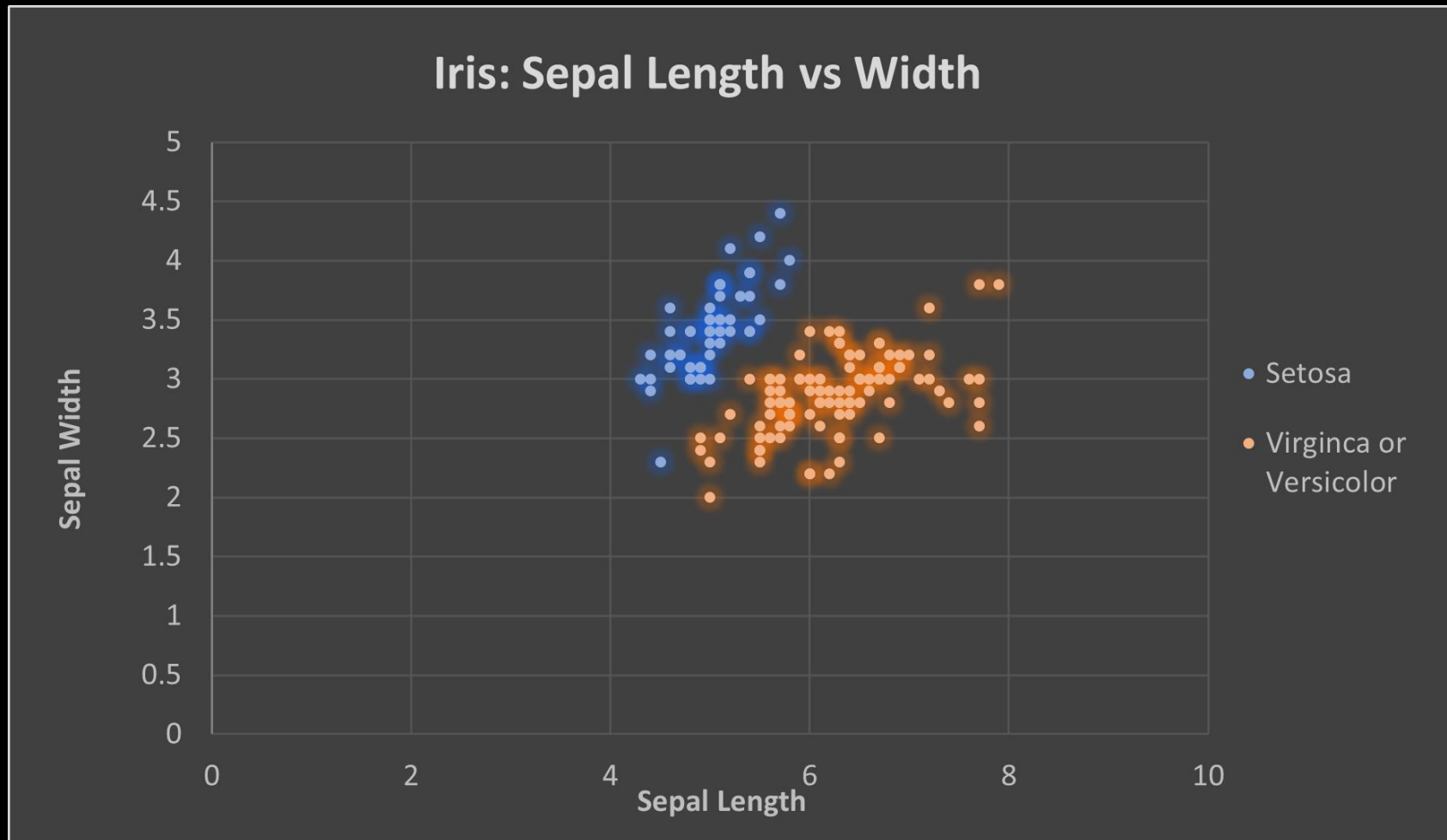
```
Epoch: 4  
Weights[0] = 0.9  
Weights[1] = 1.75  
Weights[2] = -3.25  
Weights[3] = -1.65
```

Final classifier:

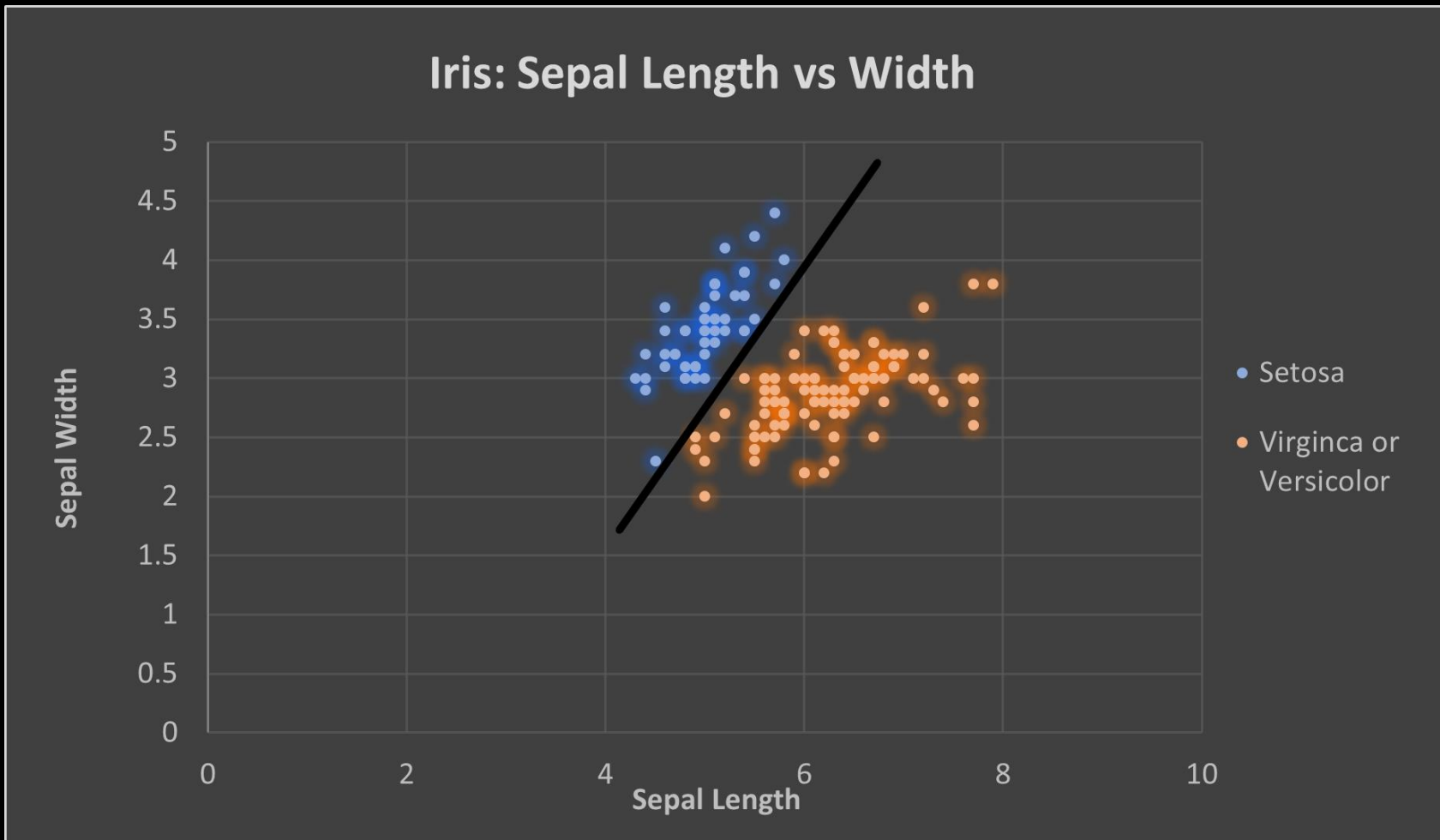
$$c(x) = 0.9x_0 + 1.75x_1 - 3.25x_2 - 1.65x_3$$



# Linearly Separable Data

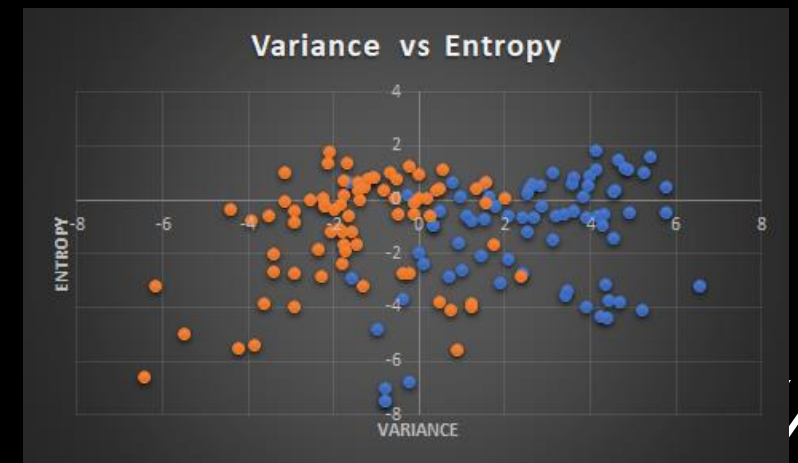
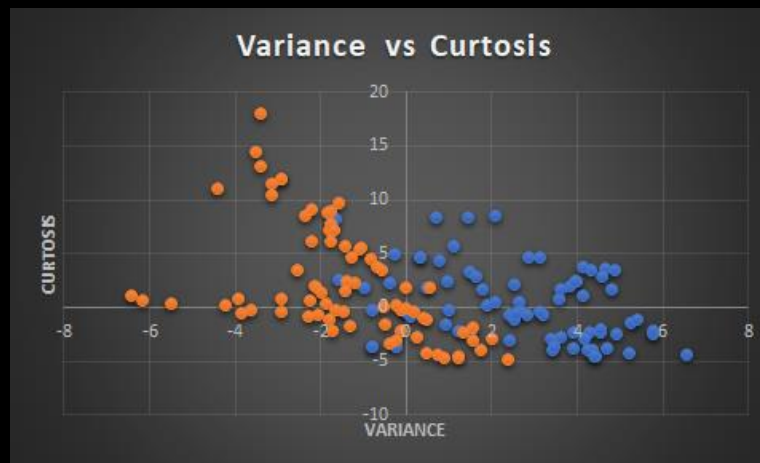
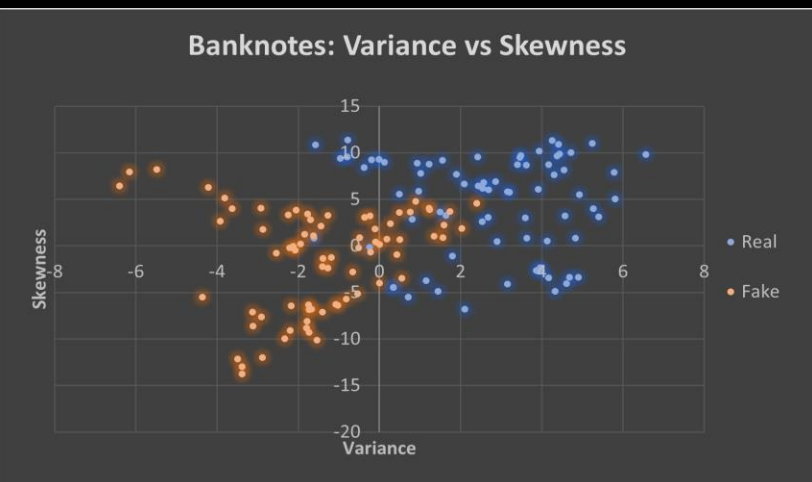
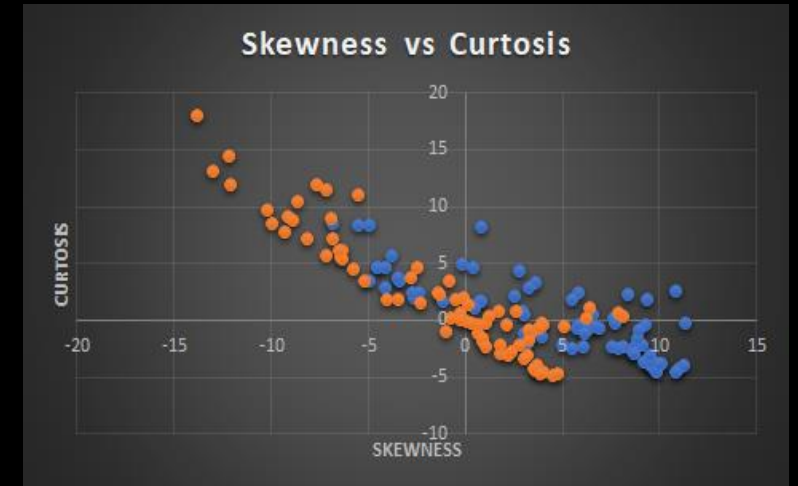
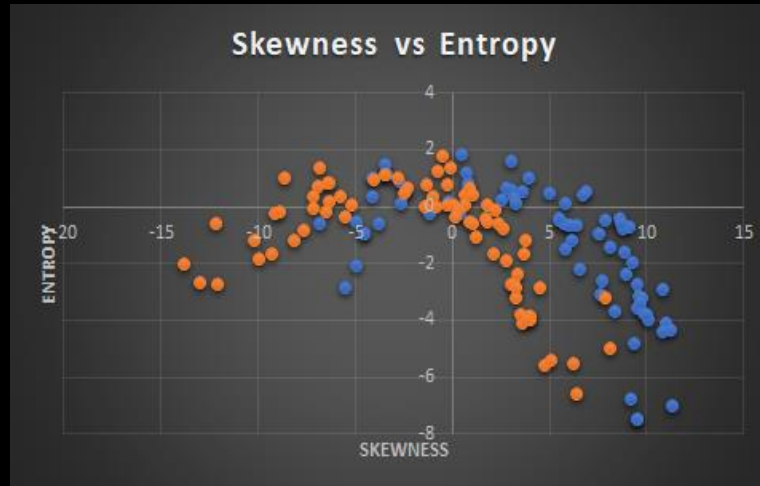
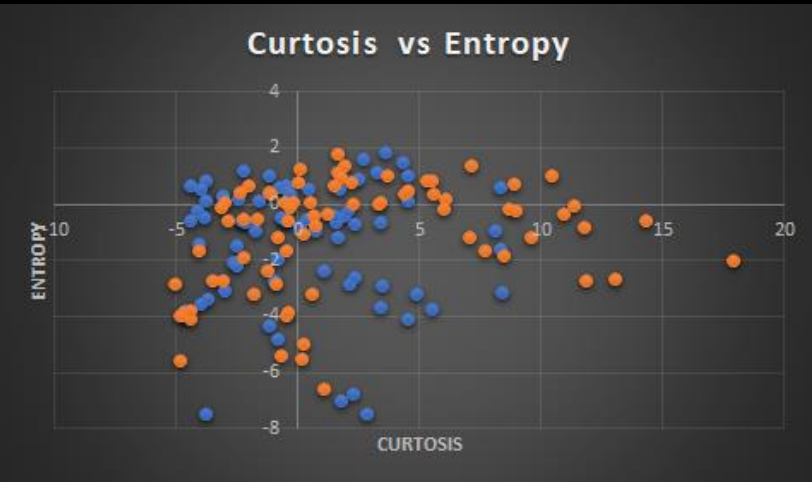


# Linearly Separable Data



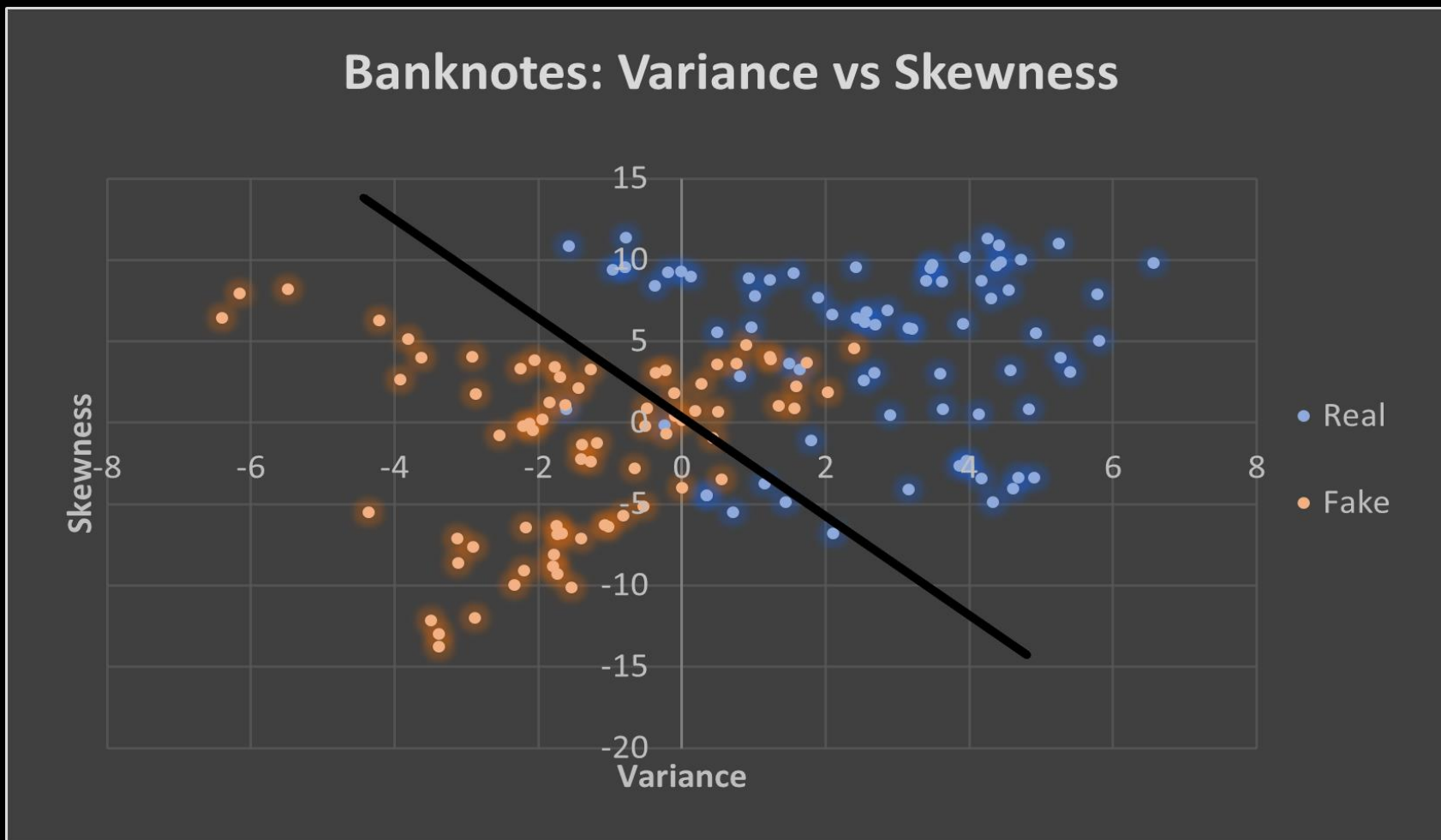


# Non-Linearly Separable Data

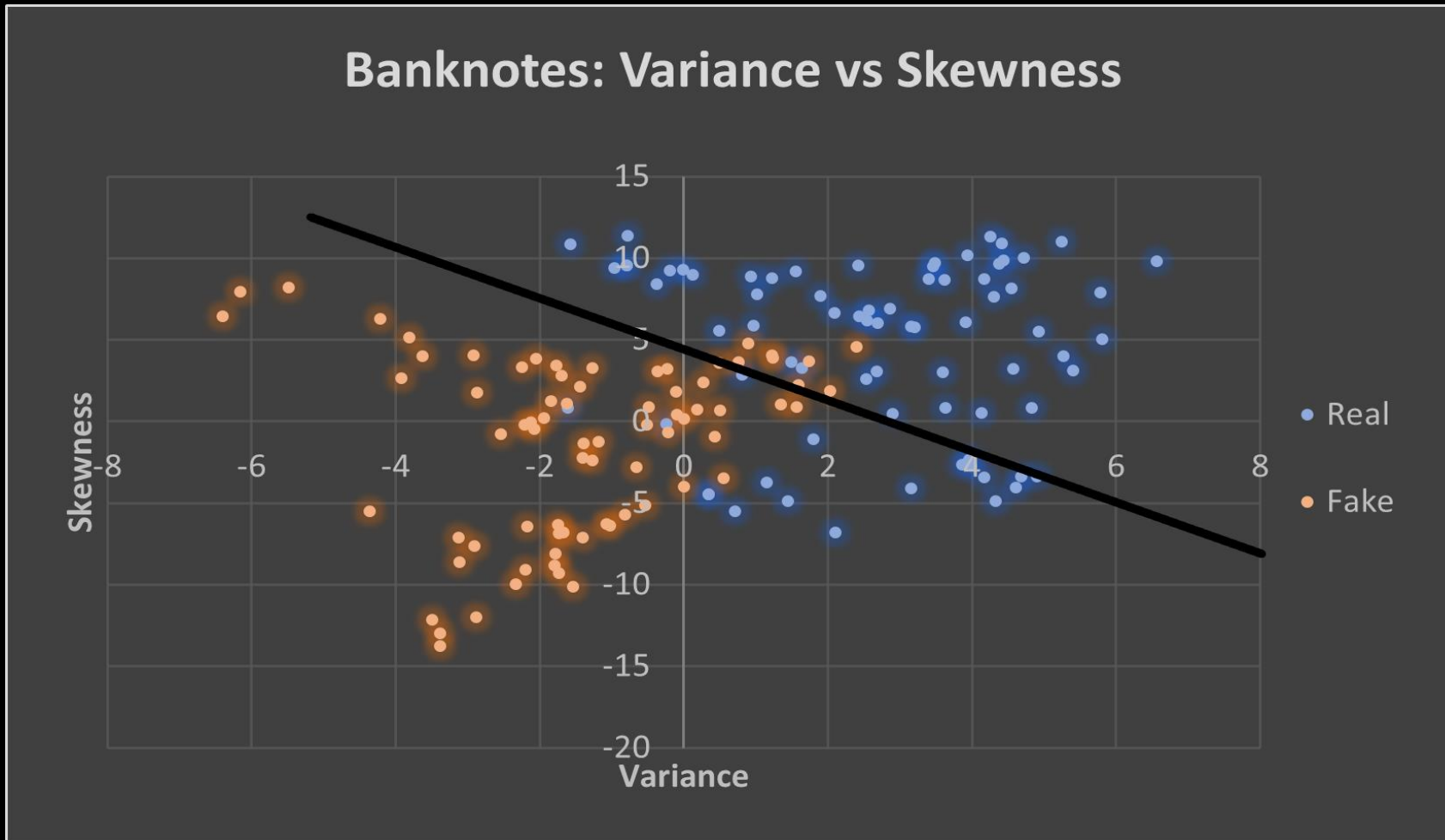


\*number of examples = 150

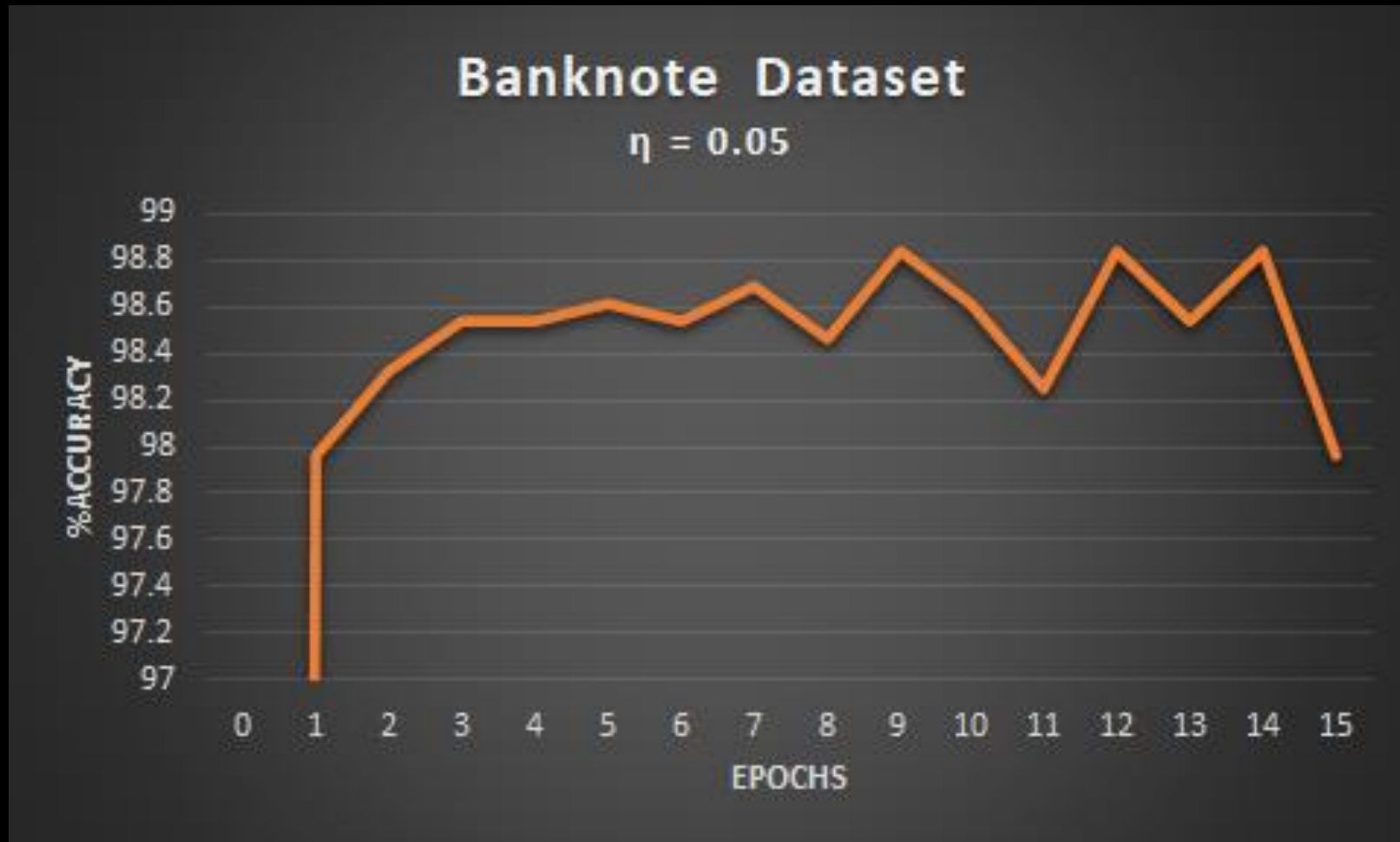
# Non-Linearly Separable Data



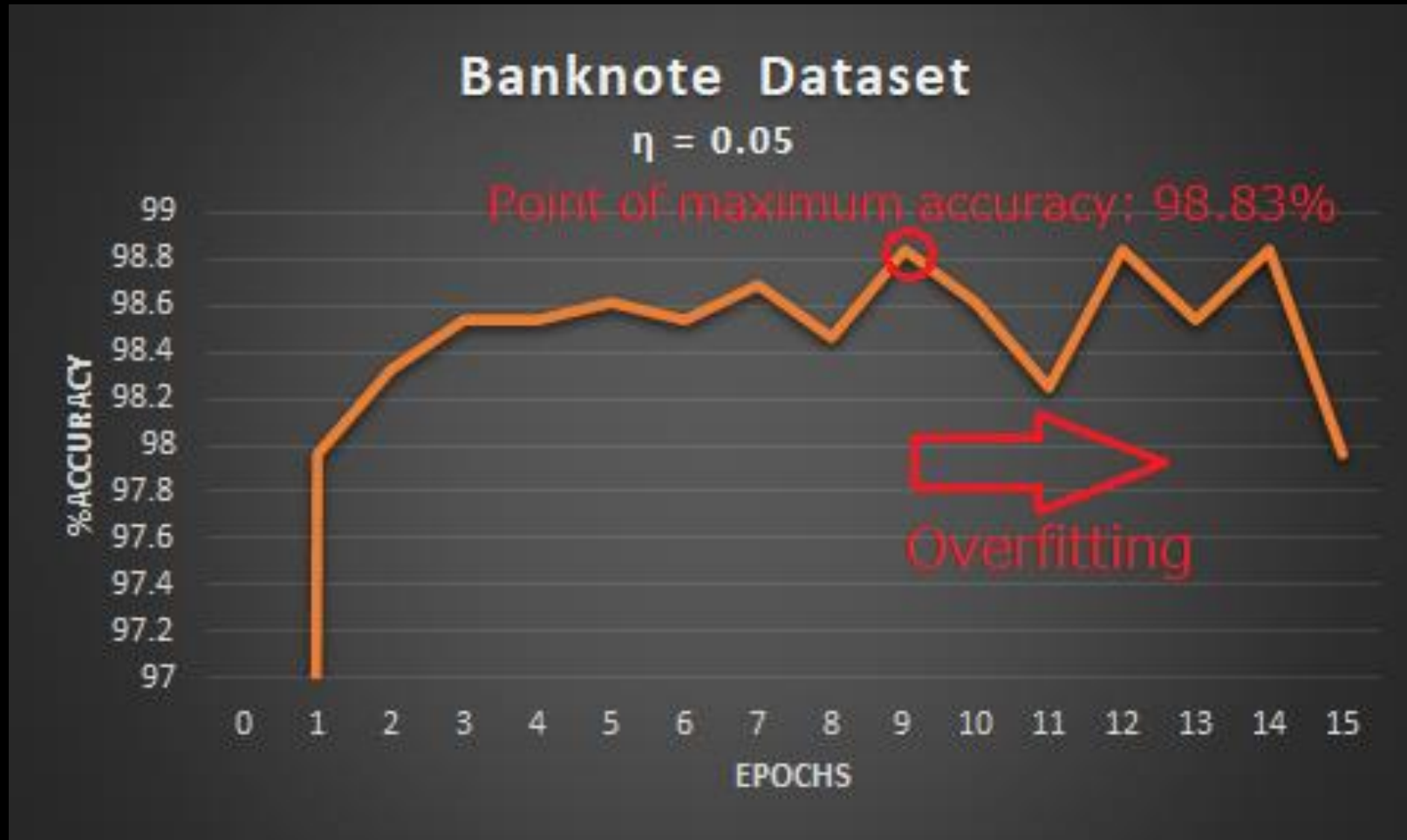
# Non-Linearly Separable Data



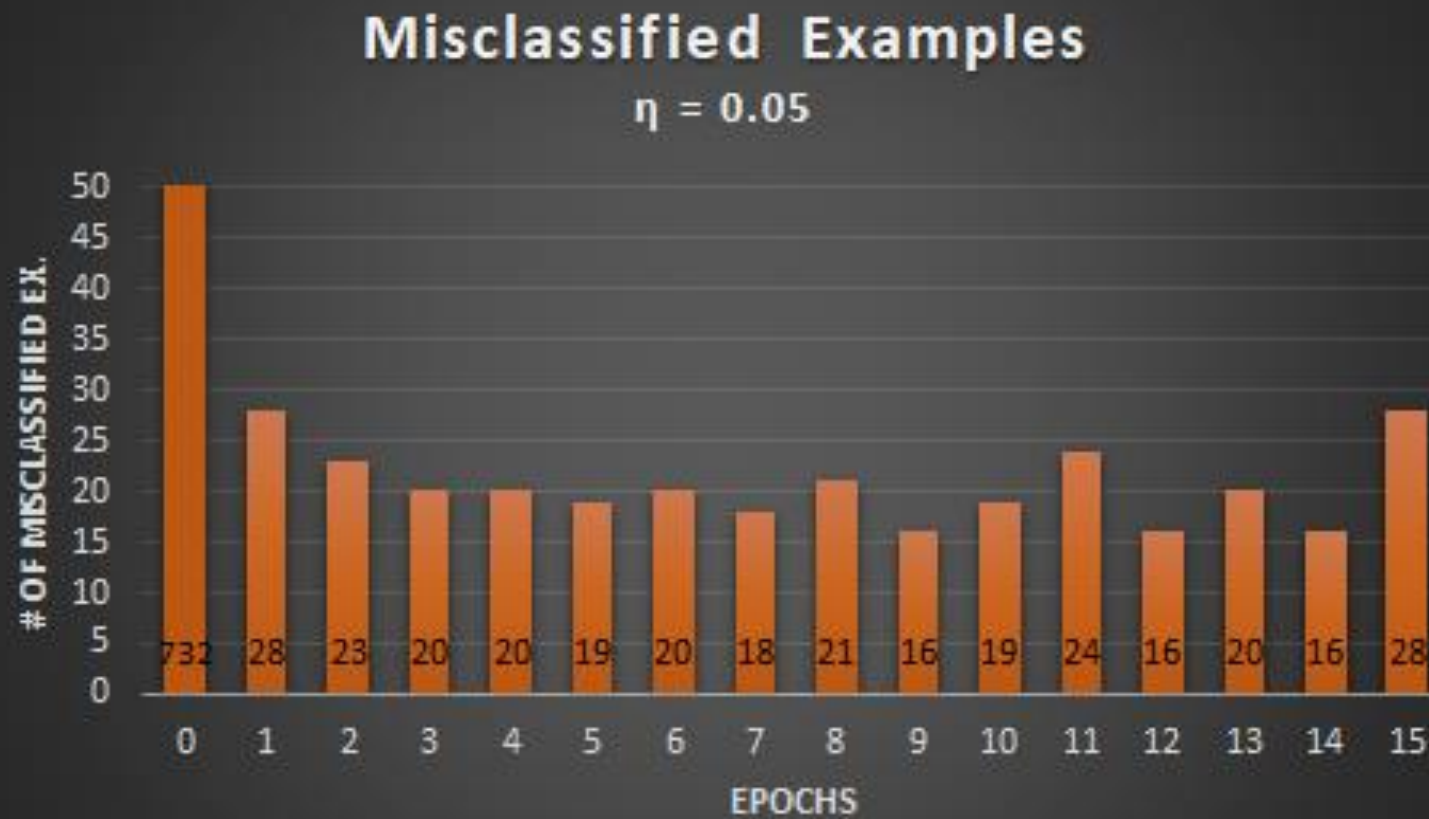
# Banknote dataset



# Banknote dataset



# Banknote dataset





# Model optimization

## Procedure:

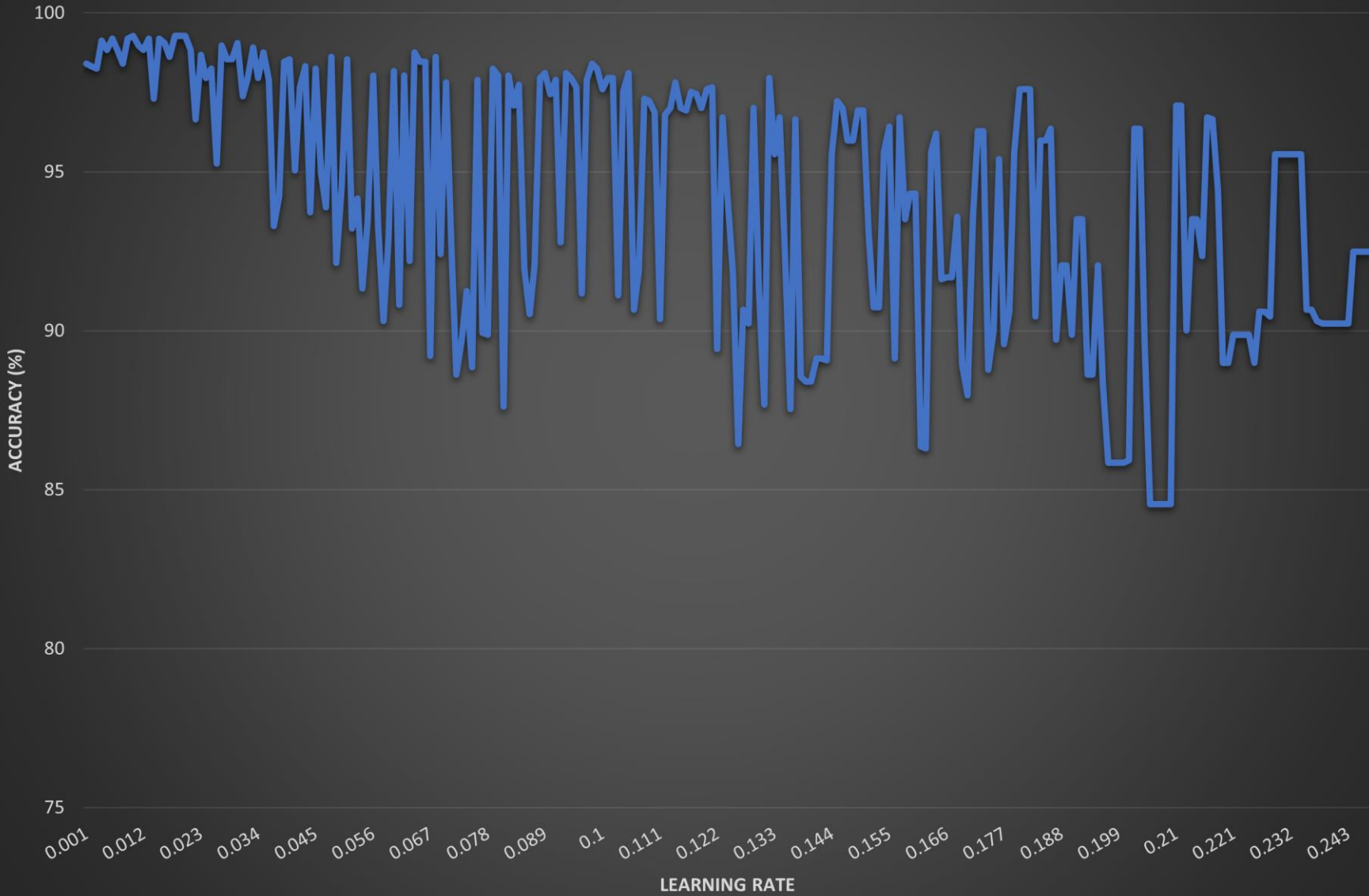
- Run through range of epochs and learning rate
  - Run the model
  - Find the point of maximum accuracy
- 
- Most useful for non-linearly separable data, as linearly separable data easily reaches 100% accuracy...



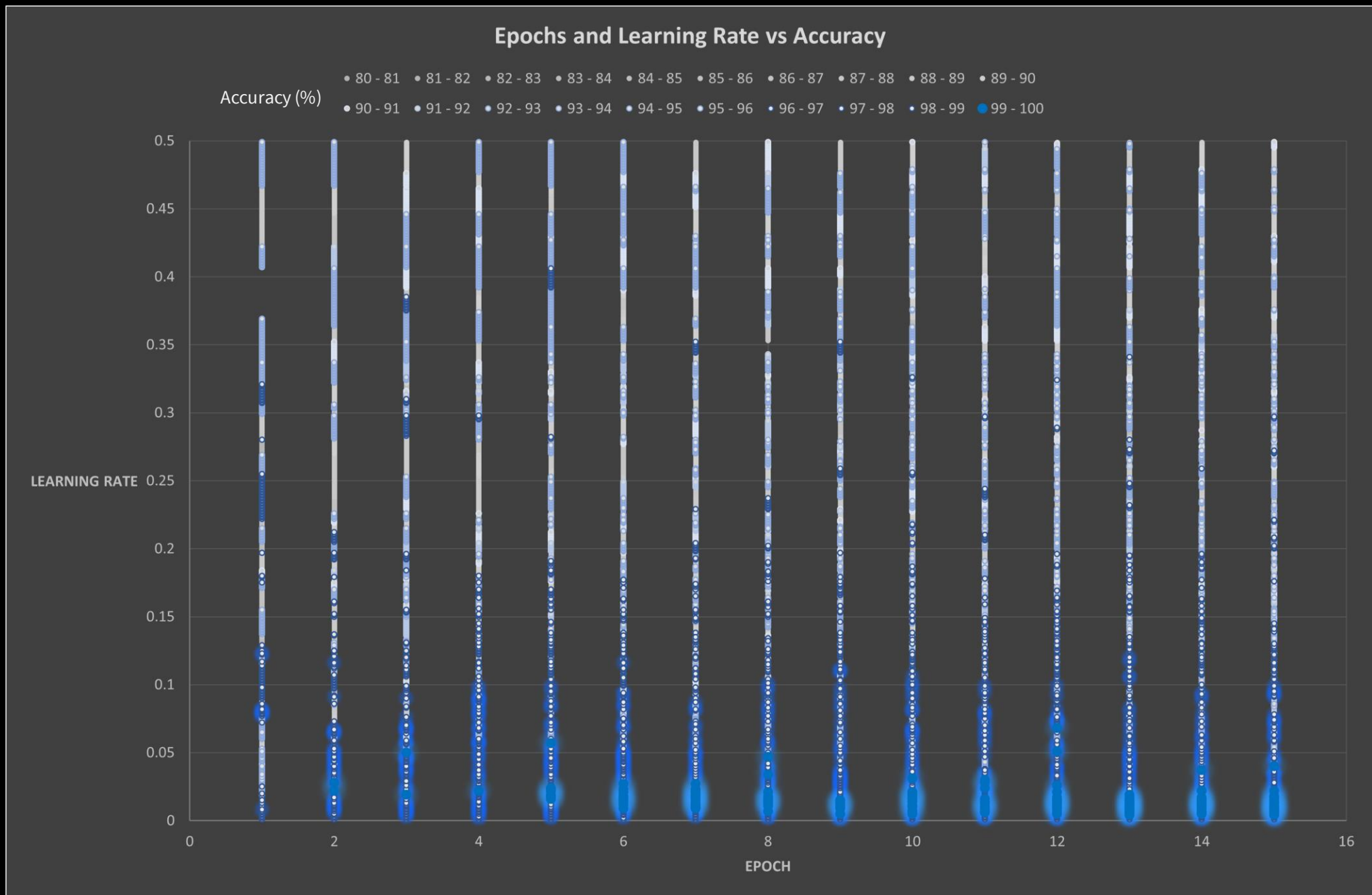


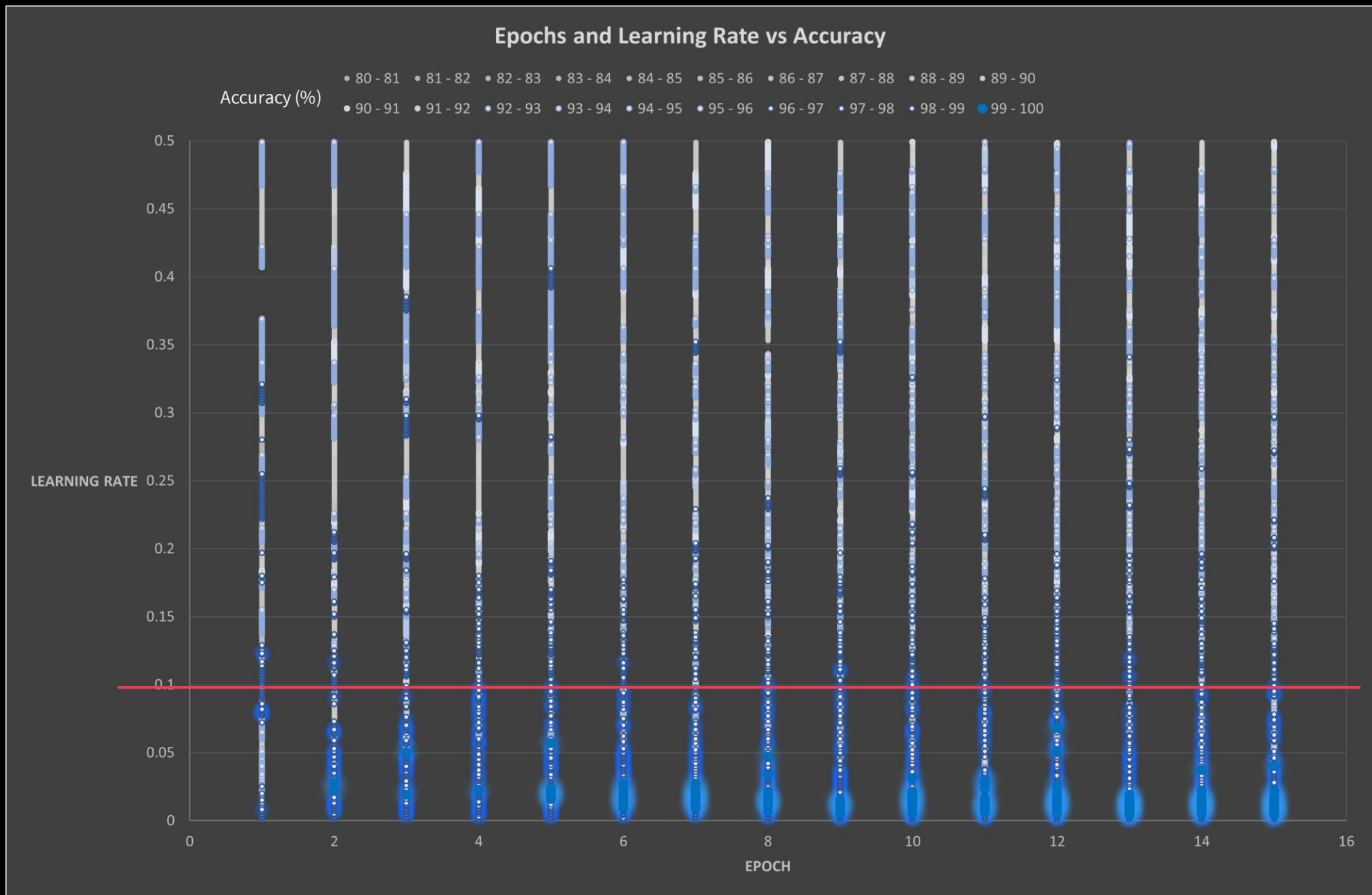


Accuracy vs Learning Rate  
Epochs = 10



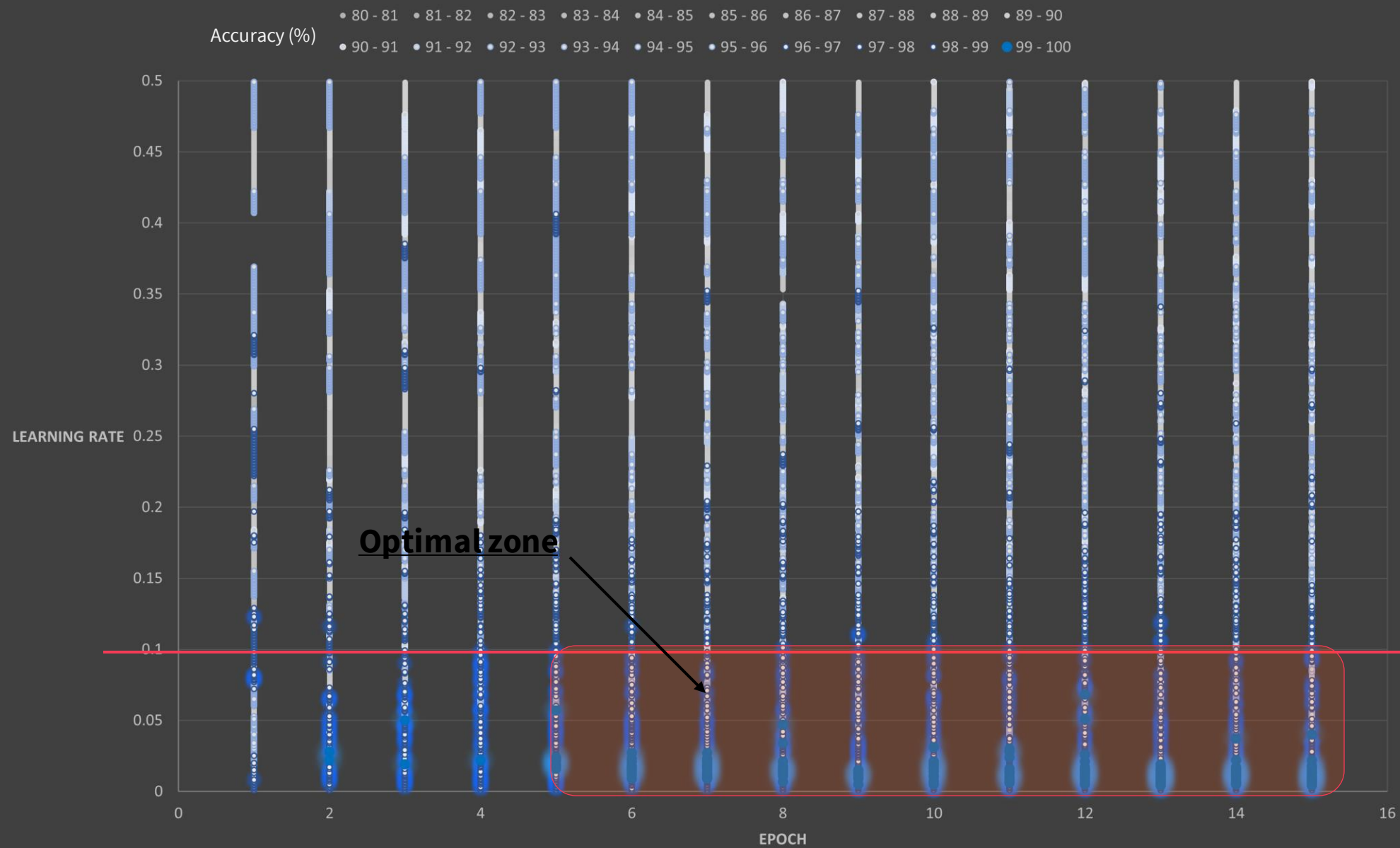








## Epochs and Learning Rate vs Accuracy





# Iris Database

- Fixed learning rate = 0.02
  - reached 100% accuracy in 5 epochs
- Optimized learning rate = 0.05
  - reached 100% accuracy in 1 epoch





# Balloons Database

- Fixed learning rate = 0.02
  - reached 100% accuracy in 5 epochs
- Optimized learning rate = 0.34
  - reached 100% accuracy in 1 epoch



# Banknote Database

- Fixed learning rate = 0.02
  - reached 99.27% accuracy in 10 epochs
- Optimized learning rate = 0.02
  - reached 99.27% accuracy in 1 epoch





# Computation costs

- Measured in epochs ( $O(N^3 + N)$  each iteration)
- Efficient for linearly separable data (~10 epochs needed)
- Less efficient for non-linearly separable data
  - How many epochs do we run?







# Disadvantages

- In order to reach 100% accuracy, data must be linearly separable
- In the real world, linearly separable data is rare
- As number of attributes increase, the data may become less linearly separable
- Irrelevant attributes (outliers):
  - Datasets with at least 1 outlier can never be linearly separable







**Thank you! :)**

