

# Privacy Preserving Decision Tree Prediction On Encrypted Data

Reem Younis, Assia Khateeb, Atheer Abo Foul, Aya Miari

Lecturer : Dr. Adi Akavia

Laboratory in Privacy Preserving Machine Learning, University of Haifa

Email: reembyounis@gmail.com, assia.khteb@gmail.com, 19aether6@gmail.com, aia-m-211@hotmail.com

August 18, 2021

## Contents

<b>1</b>	<b>Introduction</b>	<b>2</b>
<b>2</b>	<b>ARCHITECTURE</b>	<b>2</b>
<b>3</b>	<b>Single-layer Perceptron</b>	<b>2</b>
<b>4</b>	<b>Multi-layer Perceptron</b>	<b>2</b>
<b>5</b>	<b>RESULTS</b>	<b>2</b>
<b>6</b>	<b>Conclusions</b>	<b>11</b>
<b>7</b>	<b>Platforms</b>	<b>11</b>

**Abstract.** In part D we implement the prediction of iris species using perceptron neural networks trained on the Iris data set. It includes write-ups of the different types of perceptrons and their accuracy.

**Keywords**— Perceptron: either a single-layer or mutli-layer feed-forward neural network.

## 1 Introduction

Reference to our [Github repo](#).

In part D we'll describe the prediction of iris species using two different perceptron neural networks: a single-layer and multi-layer perceptron.

Each perceptron is trained and evaluated on the Iris data set split into train and test sets.

## 2 ARCHITECTURE

The main architecture of the neural networks are based on a single-layer perceptron and a multi-layer perceptron.

Each network was trained using the categorical cross entropy loss function, since the problem consists of multiple classes, and the Adam optimizer, due to its practical advantage over alternatives.

The Iris dataset was split beforehand into train and test sets. It consists of samples of 4 features, sepal length, sepal width, petal length, petal width, with 1 of 3 classes, 1.setosa, 2.versicolor, 3.virginica.

## 3 Single-layer Perceptron

The single-layer perceptron is modeled with an input layer of 4 nodes, one for each feature, and an output layer of 3 nodes, one for each class, with a softmax activation function, since the problem consists of multiple classes.

This was chosen to be a benchmark to see the performance increase with the multi-layer perceptron.

This single-layer perceptron should have a very low accuracy due to its extreme simplicity.

## 4 Multi-layer Perceptron

The multi-layer perceptron is modeled with an input layer of 4 nodes, one for each feature, 2 hidden layers of 10 nodes with a ReLU activation function, and an output layer of 3 nodes, one for each class, with a softmax activation function, since the problem consists of multiple classes. ReLU was used due to its practical advantage seen in research papers. Theoretically, the 2 hidden layers should be able to learn higher abstract information that the single-layer perceptron could not model. Thus, it is expected to produce a higher accuracy than the single-layer perceptron.

In addition to the change in the network architecture, the multi-layer perceptron also received proprocessed data, that is, data scaled down to the range of -1 to 1. This was done exclusively for the features as the classes are simply represented as a 0 or 1 and thus do not need any scaling. Theoretically, this change should allow the network to train faster as it does not need to give priority to training features that are high in value.

Finally, after each layer in the network, batch normalization was run to further allow the network to train faster and better.

## 5 RESULTS

Epoch 1/100

5/5 [=====] - 1s 0s/step - loss: 5.1102 - accuracy: 0.3258

Epoch 2/100

5/5 [=====] - 0s 4ms/step - loss: 5.0277 - accuracy: 0.3258

Epoch 3/100

5/5 [=====] - 0s 0s/step - loss: 4.9489 - accuracy: 0.3258

Epoch 4/100  
5/5 [=====] - 0s 4ms/step - loss: 4.8675 - accuracy: 0.3258  
Epoch 5/100  
5/5 [=====] - 0s 2ms/step - loss: 4.7872 - accuracy: 0.3258  
Epoch 6/100  
5/5 [=====] - 0s 0s/step - loss: 4.7102 - accuracy: 0.3258  
Epoch 7/100  
5/5 [=====] - 0s 4ms/step - loss: 4.6319 - accuracy: 0.3258  
Epoch 8/100  
5/5 [=====] - 0s 0s/step - loss: 4.5566 - accuracy: 0.3258  
Epoch 9/100  
5/5 [=====] - 0s 4ms/step - loss: 4.4817 - accuracy: 0.3258  
Epoch 10/100  
5/5 [=====] - 0s 0s/step - loss: 4.4088 - accuracy: 0.3258  
Epoch 11/100  
5/5 [=====] - 0s 4ms/step - loss: 4.3343 - accuracy: 0.3258  
Epoch 12/100  
5/5 [=====] - 0s 0s/step - loss: 4.2627 - accuracy: 0.3258  
Epoch 13/100  
5/5 [=====] - 0s 0s/step - loss: 4.1905 - accuracy: 0.3258  
Epoch 14/100  
5/5 [=====] - 0s 0s/step - loss: 4.1239 - accuracy: 0.3258  
Epoch 15/100  
5/5 [=====] - 0s 0s/step - loss: 4.0564 - accuracy: 0.3258  
Epoch 16/100  
5/5 [=====] - 0s 4ms/step - loss: 3.9927 - accuracy: 0.3258  
Epoch 17/100  
5/5 [=====] - 0s 0s/step - loss: 3.9270 - accuracy: 0.3258  
Epoch 18/100  
5/5 [=====] - 0s 4ms/step - loss: 3.8681 - accuracy: 0.3258  
Epoch 19/100  
5/5 [=====] - 0s 2ms/step - loss: 3.8133 - accuracy: 0.3258  
Epoch 20/100  
5/5 [=====] - 0s 0s/step - loss: 3.7572 - accuracy: 0.3258  
Epoch 21/100  
5/5 [=====] - 0s 0s/step - loss: 3.7059 - accuracy: 0.3258  
Epoch 22/100  
5/5 [=====] - 0s 0s/step - loss: 3.6561 - accuracy: 0.3258  
Epoch 23/100  
5/5 [=====] - 0s 0s/step - loss: 3.6096 - accuracy: 0.3258  
Epoch 24/100  
5/5 [=====] - 0s 0s/step - loss: 3.5600 - accuracy: 0.3258  
Epoch 25/100  
5/5 [=====] - 0s 0s/step - loss: 3.5156 - accuracy: 0.3258  
Epoch 26/100  
5/5 [=====] - 0s 0s/step - loss: 3.4693 - accuracy: 0.3258  
Epoch 27/100  
5/5 [=====] - 0s 0s/step - loss: 3.4267 - accuracy: 0.3258  
Epoch 28/100  
5/5 [=====] - 0s 0s/step - loss: 3.3858 - accuracy: 0.3258  
Epoch 29/100  
5/5 [=====] - 0s 0s/step - loss: 3.3485 - accuracy: 0.3258

Epoch 30/100  
5/5 [=====] - 0s 0s/step - loss: 3.3101 - accuracy: 0.3258  
Epoch 31/100  
5/5 [=====] - 0s 5ms/step - loss: 3.2757 - accuracy: 0.3258  
Epoch 32/100  
5/5 [=====] - 0s 2ms/step - loss: 3.2444 - accuracy: 0.3258  
Epoch 33/100  
5/5 [=====] - 0s 2ms/step - loss: 3.2116 - accuracy: 0.3258  
Epoch 34/100  
5/5 [=====] - 0s 2ms/step - loss: 3.1833 - accuracy: 0.3333  
Epoch 35/100  
5/5 [=====] - 0s 857us/step - loss: 3.1499 - accuracy: 0.3333  
Epoch 36/100  
5/5 [=====] - 0s 0s/step - loss: 3.1197 - accuracy: 0.3333  
Epoch 37/100  
5/5 [=====] - 0s 0s/step - loss: 3.0895 - accuracy: 0.3333  
Epoch 38/100  
5/5 [=====] - 0s 0s/step - loss: 3.0590 - accuracy: 0.3333  
Epoch 39/100  
5/5 [=====] - 0s 0s/step - loss: 3.0295 - accuracy: 0.3333  
Epoch 40/100  
5/5 [=====] - 0s 4ms/step - loss: 2.9974 - accuracy: 0.3333  
Epoch 41/100  
5/5 [=====] - 0s 0s/step - loss: 2.9681 - accuracy: 0.3333  
Epoch 42/100  
5/5 [=====] - 0s 0s/step - loss: 2.9407 - accuracy: 0.3409  
Epoch 43/100  
5/5 [=====] - 0s 0s/step - loss: 2.9120 - accuracy: 0.3409  
Epoch 44/100  
5/5 [=====] - 0s 0s/step - loss: 2.8824 - accuracy: 0.3409  
Epoch 45/100  
5/5 [=====] - 0s 0s/step - loss: 2.8530 - accuracy: 0.3409  
Epoch 46/100  
5/5 [=====] - 0s 0s/step - loss: 2.8241 - accuracy: 0.3409  
Epoch 47/100  
5/5 [=====] - 0s 0s/step - loss: 2.7919 - accuracy: 0.3485  
Epoch 48/100  
5/5 [=====] - 0s 4ms/step - loss: 2.7636 - accuracy: 0.3485  
Epoch 49/100  
5/5 [=====] - 0s 0s/step - loss: 2.7353 - accuracy: 0.3561  
Epoch 50/100  
5/5 [=====] - 0s 0s/step - loss: 2.7091 - accuracy: 0.3561  
Epoch 51/100  
5/5 [=====] - 0s 0s/step - loss: 2.6801 - accuracy: 0.3561  
Epoch 52/100  
5/5 [=====] - 0s 0s/step - loss: 2.6494 - accuracy: 0.3561  
Epoch 53/100  
5/5 [=====] - 0s 0s/step - loss: 2.6184 - accuracy: 0.3561  
Epoch 54/100  
5/5 [=====] - 0s 2ms/step - loss: 2.5881 - accuracy: 0.3561  
Epoch 55/100  
5/5 [=====] - 0s 0s/step - loss: 2.5570 - accuracy: 0.3561

Epoch 56/100  
5/5 [=====] - 0s 4ms/step - loss: 2.5264 - accuracy: 0.3636  
Epoch 57/100  
5/5 [=====] - 0s 0s/step - loss: 2.4954 - accuracy: 0.3712  
Epoch 58/100  
5/5 [=====] - 0s 0s/step - loss: 2.4648 - accuracy: 0.3561  
Epoch 59/100  
5/5 [=====] - 0s 0s/step - loss: 2.4381 - accuracy: 0.3561  
Epoch 60/100  
5/5 [=====] - 0s 0s/step - loss: 2.4102 - accuracy: 0.3561  
Epoch 61/100  
5/5 [=====] - 0s 4ms/step - loss: 2.3837 - accuracy: 0.3636  
Epoch 62/100  
5/5 [=====] - 0s 0s/step - loss: 2.3581 - accuracy: 0.3712  
Epoch 63/100  
5/5 [=====] - 0s 4ms/step - loss: 2.3293 - accuracy: 0.3712  
Epoch 64/100  
5/5 [=====] - 0s 0s/step - loss: 2.3017 - accuracy: 0.3712  
Epoch 65/100  
5/5 [=====] - 0s 4ms/step - loss: 2.2719 - accuracy: 0.3712  
Epoch 66/100  
5/5 [=====] - 0s 0s/step - loss: 2.2419 - accuracy: 0.3864  
Epoch 67/100  
5/5 [=====] - 0s 0s/step - loss: 2.2144 - accuracy: 0.3864  
Epoch 68/100  
5/5 [=====] - 0s 0s/step - loss: 2.1865 - accuracy: 0.3864  
Epoch 69/100  
5/5 [=====] - 0s 0s/step - loss: 2.1579 - accuracy: 0.3712  
Epoch 70/100  
5/5 [=====] - 0s 4ms/step - loss: 2.1304 - accuracy: 0.3712  
Epoch 71/100  
5/5 [=====] - 0s 0s/step - loss: 2.1048 - accuracy: 0.3636  
Epoch 72/100  
5/5 [=====] - 0s 4ms/step - loss: 2.0768 - accuracy: 0.3561  
Epoch 73/100  
5/5 [=====] - 0s 0s/step - loss: 2.0452 - accuracy: 0.3561  
Epoch 74/100  
5/5 [=====] - 0s 0s/step - loss: 2.0167 - accuracy: 0.3636  
Epoch 75/100  
5/5 [=====] - 0s 0s/step - loss: 1.9872 - accuracy: 0.3636  
Epoch 76/100  
5/5 [=====] - 0s 0s/step - loss: 1.9575 - accuracy: 0.3636  
Epoch 77/100  
5/5 [=====] - 0s 4ms/step - loss: 1.9277 - accuracy: 0.3864  
Epoch 78/100  
5/5 [=====] - 0s 0s/step - loss: 1.8979 - accuracy: 0.3864  
Epoch 79/100  
5/5 [=====] - 0s 4ms/step - loss: 1.8672 - accuracy: 0.3864  
Epoch 80/100  
5/5 [=====] - 0s 2ms/step - loss: 1.8378 - accuracy: 0.3864  
Epoch 81/100  
5/5 [=====] - 0s 0s/step - loss: 1.8088 - accuracy: 0.3864

Epoch 82/100  
5/5 [=====] - 0s 0s/step - loss: 1.7809 - accuracy: 0.3864  
Epoch 83/100  
5/5 [=====] - 0s 0s/step - loss: 1.7501 - accuracy: 0.3864  
Epoch 84/100  
5/5 [=====] - 0s 4ms/step - loss: 1.7235 - accuracy: 0.4091  
Epoch 85/100  
5/5 [=====] - 0s 0s/step - loss: 1.6962 - accuracy: 0.4167  
Epoch 86/100  
5/5 [=====] - 0s 4ms/step - loss: 1.6689 - accuracy: 0.4242  
Epoch 87/100  
5/5 [=====] - 0s 0s/step - loss: 1.6422 - accuracy: 0.4394  
Epoch 88/100  
5/5 [=====] - 0s 4ms/step - loss: 1.6151 - accuracy: 0.4470  
Epoch 89/100  
5/5 [=====] - 0s 0s/step - loss: 1.5876 - accuracy: 0.4470  
Epoch 90/100  
5/5 [=====] - 0s 4ms/step - loss: 1.5601 - accuracy: 0.4621  
Epoch 91/100  
5/5 [=====] - 0s 0s/step - loss: 1.5331 - accuracy: 0.4621  
Epoch 92/100  
5/5 [=====] - 0s 2ms/step - loss: 1.5058 - accuracy: 0.4621  
Epoch 93/100  
5/5 [=====] - 0s 0s/step - loss: 1.4769 - accuracy: 0.4470  
Epoch 94/100  
5/5 [=====] - 0s 0s/step - loss: 1.4504 - accuracy: 0.4242  
Epoch 95/100  
5/5 [=====] - 0s 0s/step - loss: 1.4230 - accuracy: 0.4242  
Epoch 96/100  
5/5 [=====] - 0s 0s/step - loss: 1.3961 - accuracy: 0.4242  
Epoch 97/100  
5/5 [=====] - 0s 1ms/step - loss: 1.3708 - accuracy: 0.4318  
Epoch 98/100  
5/5 [=====] - 0s 998us/step - loss: 1.3444 - accuracy: 0.4470  
Epoch 99/100  
5/5 [=====] - 0s 997us/step - loss: 1.3182 - accuracy: 0.4545  
Epoch 100/100  
5/5 [=====] - 0s 1ms/step - loss: 1.2951 - accuracy: 0.4545  
1/1 [=====] - 0s 213ms/step - loss: 0.8540 - accuracy: 0.6667  
Epoch 1/100  
5/5 [=====] - 1s 4ms/step - loss: 1.8799 - accuracy: 0.2500  
Epoch 2/100  
5/5 [=====] - 0s 4ms/step - loss: 1.6343 - accuracy: 0.3258  
Epoch 3/100  
5/5 [=====] - 0s 4ms/step - loss: 1.5378 - accuracy: 0.3712  
Epoch 4/100  
5/5 [=====] - 0s 0s/step - loss: 1.4903 - accuracy: 0.4242  
Epoch 5/100  
5/5 [=====] - 0s 4ms/step - loss: 1.3035 - accuracy: 0.4773  
Epoch 6/100  
5/5 [=====] - 0s 4ms/step - loss: 1.2276 - accuracy: 0.5379  
Epoch 7/100

5/5 [=====] - 0s 0s/step - loss: 1.1522 - accuracy: 0.5379  
 Epoch 8/100  
 5/5 [=====] - 0s 4ms/step - loss: 1.0671 - accuracy: 0.5758  
 Epoch 9/100  
 5/5 [=====] - 0s 0s/step - loss: 1.0572 - accuracy: 0.5379  
 Epoch 10/100  
 5/5 [=====] - 0s 5ms/step - loss: 0.9462 - accuracy: 0.5985  
 Epoch 11/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.9056 - accuracy: 0.5758  
 Epoch 12/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.8131 - accuracy: 0.5985  
 Epoch 13/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.8191 - accuracy: 0.6136  
 Epoch 14/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.7529 - accuracy: 0.6667  
 Epoch 15/100  
 5/5 [=====] - 0s 262us/step - loss: 0.6988 - accuracy: 0.6667  
 Epoch 16/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.6868 - accuracy: 0.6515  
 Epoch 17/100  
 5/5 [=====] - 0s 0s/step - loss: 0.6499 - accuracy: 0.6894  
 Epoch 18/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.5962 - accuracy: 0.7197  
 Epoch 19/100  
 5/5 [=====] - 0s 0s/step - loss: 0.5714 - accuracy: 0.7424  
 Epoch 20/100  
 5/5 [=====] - 0s 0s/step - loss: 0.5506 - accuracy: 0.7576  
 Epoch 21/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.5127 - accuracy: 0.8258  
 Epoch 22/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.5111 - accuracy: 0.8106  
 Epoch 23/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.4799 - accuracy: 0.8333  
 Epoch 24/100  
 5/5 [=====] - 0s 0s/step - loss: 0.4614 - accuracy: 0.8485  
 Epoch 25/100  
 5/5 [=====] - 0s 0s/step - loss: 0.4414 - accuracy: 0.8485  
 Epoch 26/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.4581 - accuracy: 0.8409  
 Epoch 27/100  
 5/5 [=====] - 0s 0s/step - loss: 0.4167 - accuracy: 0.8485  
 Epoch 28/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.4099 - accuracy: 0.8561  
 Epoch 29/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.4126 - accuracy: 0.8712  
 Epoch 30/100  
 5/5 [=====] - 0s 0s/step - loss: 0.3506 - accuracy: 0.8939  
 Epoch 31/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.3735 - accuracy: 0.9015  
 Epoch 32/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.3579 - accuracy: 0.8939  
 Epoch 33/100

5/5 [=====] - 0s 0s/step - loss: 0.3769 - accuracy: 0.8939  
 Epoch 34/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.3269 - accuracy: 0.9167  
 Epoch 35/100  
 5/5 [=====] - 0s 0s/step - loss: 0.3213 - accuracy: 0.9091  
 Epoch 36/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.3181 - accuracy: 0.9091  
 Epoch 37/100  
 5/5 [=====] - 0s 0s/step - loss: 0.3084 - accuracy: 0.9318  
 Epoch 38/100  
 5/5 [=====] - 0s 0s/step - loss: 0.3253 - accuracy: 0.9242  
 Epoch 39/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.3335 - accuracy: 0.9167  
 Epoch 40/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2868 - accuracy: 0.9167  
 Epoch 41/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2752 - accuracy: 0.9394  
 Epoch 42/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2774 - accuracy: 0.9470  
 Epoch 43/100  
 5/5 [=====] - 0s 6ms/step - loss: 0.3153 - accuracy: 0.9091  
 Epoch 44/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2885 - accuracy: 0.9167  
 Epoch 45/100  
 5/5 [=====] - 0s 0s/step - loss: 0.3074 - accuracy: 0.9318  
 Epoch 46/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2612 - accuracy: 0.9394  
 Epoch 47/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2433 - accuracy: 0.9394  
 Epoch 48/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2732 - accuracy: 0.9242  
 Epoch 49/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2604 - accuracy: 0.9167  
 Epoch 50/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2491 - accuracy: 0.9318  
 Epoch 51/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2314 - accuracy: 0.9470  
 Epoch 52/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.2539 - accuracy: 0.9394  
 Epoch 53/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2223 - accuracy: 0.9242  
 Epoch 54/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2067 - accuracy: 0.9545  
 Epoch 55/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2257 - accuracy: 0.9394  
 Epoch 56/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1948 - accuracy: 0.9697  
 Epoch 57/100  
 5/5 [=====] - 0s 5ms/step - loss: 0.2136 - accuracy: 0.9545  
 Epoch 58/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.2149 - accuracy: 0.9621  
 Epoch 59/100



5/5 [=====] - 0s 2ms/step - loss: 0.1907 - accuracy: 0.9621  
 Epoch 60/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.1968 - accuracy: 0.9545  
 Epoch 61/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.1941 - accuracy: 0.9394  
 Epoch 62/100  
 5/5 [=====] - 0s 0s/step - loss: 0.2294 - accuracy: 0.9318  
 Epoch 63/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.2045 - accuracy: 0.9242  
 Epoch 64/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1784 - accuracy: 0.9545  
 Epoch 65/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1902 - accuracy: 0.9470  
 Epoch 66/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1882 - accuracy: 0.9470  
 Epoch 67/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1751 - accuracy: 0.9621  
 Epoch 68/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1750 - accuracy: 0.9394  
 Epoch 69/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1843 - accuracy: 0.9470  
 Epoch 70/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1715 - accuracy: 0.9545  
 Epoch 71/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.2200 - accuracy: 0.9242  
 Epoch 72/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1641 - accuracy: 0.9697  
 Epoch 73/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1988 - accuracy: 0.9470  
 Epoch 74/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1971 - accuracy: 0.9470  
 Epoch 75/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1560 - accuracy: 0.9621  
 Epoch 76/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1892 - accuracy: 0.9470  
 Epoch 77/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1928 - accuracy: 0.9545  
 Epoch 78/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1630 - accuracy: 0.9697  
 Epoch 79/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1838 - accuracy: 0.9545  
 Epoch 80/100  
 5/5 [=====] - 0s 6ms/step - loss: 0.1537 - accuracy: 0.9621  
 Epoch 81/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1483 - accuracy: 0.9697  
 Epoch 82/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1469 - accuracy: 0.9621  
 Epoch 83/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1927 - accuracy: 0.9470  
 Epoch 84/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1749 - accuracy: 0.9545  
 Epoch 85/100

5/5 [=====] - 0s 0s/step - loss: 0.1645 - accuracy: 0.9545  
 Epoch 86/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1411 - accuracy: 0.9545  
 Epoch 87/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1450 - accuracy: 0.9773  
 Epoch 88/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1746 - accuracy: 0.9545  
 Epoch 89/100  
 5/5 [=====] - 0s 2ms/step - loss: 0.1279 - accuracy: 0.9697  
 Epoch 90/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1316 - accuracy: 0.9848  
 Epoch 91/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1670 - accuracy: 0.9470  
 Epoch 92/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1805 - accuracy: 0.9318  
 Epoch 93/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1345 - accuracy: 0.9621  
 Epoch 94/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1790 - accuracy: 0.9545  
 Epoch 95/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1490 - accuracy: 0.9621  
 Epoch 96/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1435 - accuracy: 0.9394  
 Epoch 97/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1634 - accuracy: 0.9470  
 Epoch 98/100  
 5/5 [=====] - 0s 0s/step - loss: 0.1682 - accuracy: 0.9470  
 Epoch 99/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1984 - accuracy: 0.9167  
 Epoch 100/100  
 5/5 [=====] - 0s 4ms/step - loss: 0.1511 - accuracy: 0.9470  
 1/1 [=====] - 0s 250ms/step - loss: 0.2106 - accuracy: 0.9333

Naive Accuracy: 0.6666666865348816

Naive Predictions:

```

[[0.01354977 0.5061691 0.48028105]
[0.06246578 0.56894875 0.36858547]
[0.0074184 0.44270876 0.5498728 ]
[0.01080847 0.4541977 0.53499377]
[0.02312316 0.55372 0.4231569 ]
[0.00527738 0.39130107 0.60342157]
[0.00576249 0.52103436 0.4732032 ]
[0.00304265 0.31141222 0.68554515]
[0.07634713 0.58648854 0.33716434]
[0.00298612 0.3317028 0.6653111 ]
[0.00340858 0.28069374 0.7158977 ]
[0.01619304 0.5350266 0.44878045]
[0.00580444 0.32919884 0.66499674]
[0.01750648 0.56174 0.4207535 ]
[0.00643218 0.49766505 0.49590284]]
  
```

Better Accuracy: 0.9333333373069763

Better Predictions:

```
[[0.00862633 0.9620326 0.02934105]
[0.9790445 0.01505063 0.00590489]
[0.00578253 0.9523178 0.0418997 ]
[0.00888783 0.97394866 0.01716346]
[0.07783689 0.7403439 0.18181916]
[0.01994696 0.21152395 0.7685291 ]
[0.0148526 0.86186147 0.12328597]
[0.00590803 0.02657974 0.96751225]
[0.9846364 0.0105518 0.00481179]
[0.00601991 0.05104782 0.94293225]
[0.00143864 0.0452151 0.95334625]
[0.01584265 0.9249697 0.0591877 ]
[0.01459309 0.3523831 0.63302374]
[0.01764467 0.9287313 0.05362389]
[0.01701402 0.77023745 0.21274848]]
```

As seen in the example run, the multi-layer perceptron performed much better than the single-layer perceptron. The accuracy for the single-layer perceptron and multi-layer perceptron was 66.666% and 93.333%, respectively.

Thus, it can easily be stated that the hidden layers and preprocessing of data allowed the multi-layer perceptron to much better learn the data.

The 2 hidden layers allowed the multi-layer perceptron to learn higher abstract features about the Iris data set and thus produced a higher accuracy. Furthermore, the data scaling and batch normalization increased the network's learning rate.

Although a 93.333% accuracy is astounding, this is not likely to continue if the test data set size increased.

## 6 Conclusions

Overall, the results gathered follow the expectation/hypothesis, that is, that a more complex network, but not too complex, would be able to learn much more effectively than a simple one. This is likely in many scenarios.

However, it is likely that the bias-variance tradeoff effect would come into play here. That is, as the bias of a model increases, its variance will as well, and vice-versa. A simple model has extreme bias, but no variance.

A extremely complex model has high variance, but almost no bias. Thus, an extremely complex model is likely to overfit a data set, while a simple model is likely to underfit a data set.

Therefore, it is evident that caution must be taken to find a point where the bias and variance are minimized while still producing an effective model.

This can be done by using a technique like cross-validation to produce models that are each trained and validated against different "folds" of the train data set. Then, the model that minimizes the loss function (against the sum of error of the validation folds) can be used as the final model and can be run on the test set to produce the resulting accuracy.

## 7 Platforms

- pycharm, python 3.9
- overleaf.com