

Traditional Machine Learning Method vs Neural Networks (on the MNIST Handwritten Dataset)

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Abstract— In this paper we compare the accuracies of solving the task of number identification of the MNIST handwritten dataset with two types of modelling approaches. On the one hand, we use well known traditional machine learning classification which is Logistic Regression in this case; and on the other hand we use Artificial Neural Networks to do the same. Traditional Machine Learning algorithms tend to perform at the same level when the data size increases but ANN outperforms traditional Machine Learning algorithms. Three different groups of models are trained. For the entire dataset, to detect '2' and to detect 'not 2'. In addition we add another hidden layer to the neural networks to offer more insight.

I. INTRODUCTION

The MNIST database (Modified National Institute of Standards and Technology database) is a large database of handwritten digits that is commonly used for training various image processing systems. The database is also widely used for training and testing in the field of machine learning. It was created by "re-mixing" the samples from NIST's original datasets. The creators felt that since NIST's training dataset was taken from American Census Bureau employees, while the testing dataset was taken from American high school students, it was not well-suited for machine learning experiments. Furthermore, the black and white images from NIST were normalized to fit into a 28x28 pixel bounding box and anti-aliased, which introduced grayscale levels.

The MNIST database contains 60,000 training images and 10,000 testing images. Half of the training set and half of the test set were taken from NIST's training dataset, while the other half of the training set and the other half of the test set were taken from NIST's testing dataset.

II. METHOD

A. Methodology Overview

The MNIST data is freely available online. It is split into 60,000 training images and 10,000 testing images. The data was flattened and shuffled before it was used for both the Logistic Regression classifier as well as the Neural network. The dataset was also shuffled before being used to train the two models. The code was written using Sklearn, Tensorflow and Keras in Python

In order to gain a better insight into the difference between the methods, three different sets of models were trained for the dataset. One set was a general model for the entire dataset, the second was to identify just the number '2' and the third was a model to identify 'not 2'. In addition, another neural network with a hidden layer was trained in each set.

The sets are as follows-

1. Set 1 -
 - a. Logistic Regression classifier on the entire dataset
 - b. Neural Network trained on the entire dataset
2. Set 2 -
 - a. Logistic Regression classifier to find '2'
 - b. Neural Network trained to identify '2'
3. Set 3 -
 - a. Logistic Regression classifier to find 'not 2'
 - b. Neural Network trained to find 'not 2'

Additionally, each model had K-fold validation done on it with three folds. To evaluate the performance of any machine learning model we need to test it on some unseen data. Based on the models' performance on unseen data we can say whether our model is Under-fitting/Over-fitting/Well generalized. This is the reason we have used K-fold cross-validation.

Once the models are trained we then find out their accuracy and find the confusion matrix along with our performance metrics such as precision, recall and f1 score. The same was done for the cross-validation score and cross-validation predictions.

B. Logistic Regression Model

The logistic regression classifier was trained using the 'lbfgs' solver and a tolerance of 0.1.

C. Neural Network Model

The neural networks were trained using the 'sigmoid' activation function, the 'adam' optimizer and the 'sparse_categorical_crossentropy' loss function and five epochs. They had 10 dense units/neurons.

The neural network with the hidden layer had an additional layer with 100 dense units using the 'relu' activation function.

III. RESULTS

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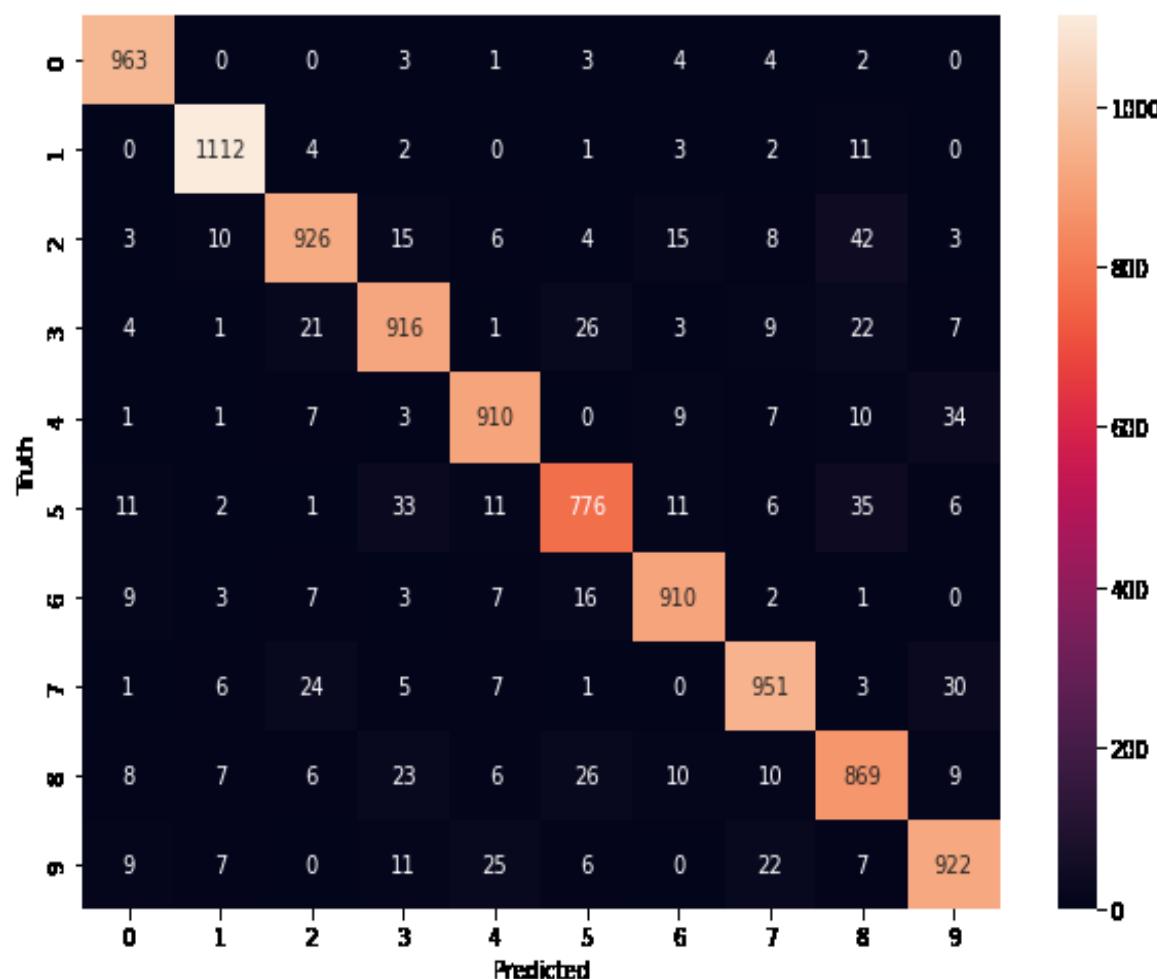
Finally, complete content and organizational editing before formatting. Please take note of the following items when proofreading spelling and grammar:

A. Results for the entire dataset

Regression Model

Accuracy = 0.9255

Confusion Matrix



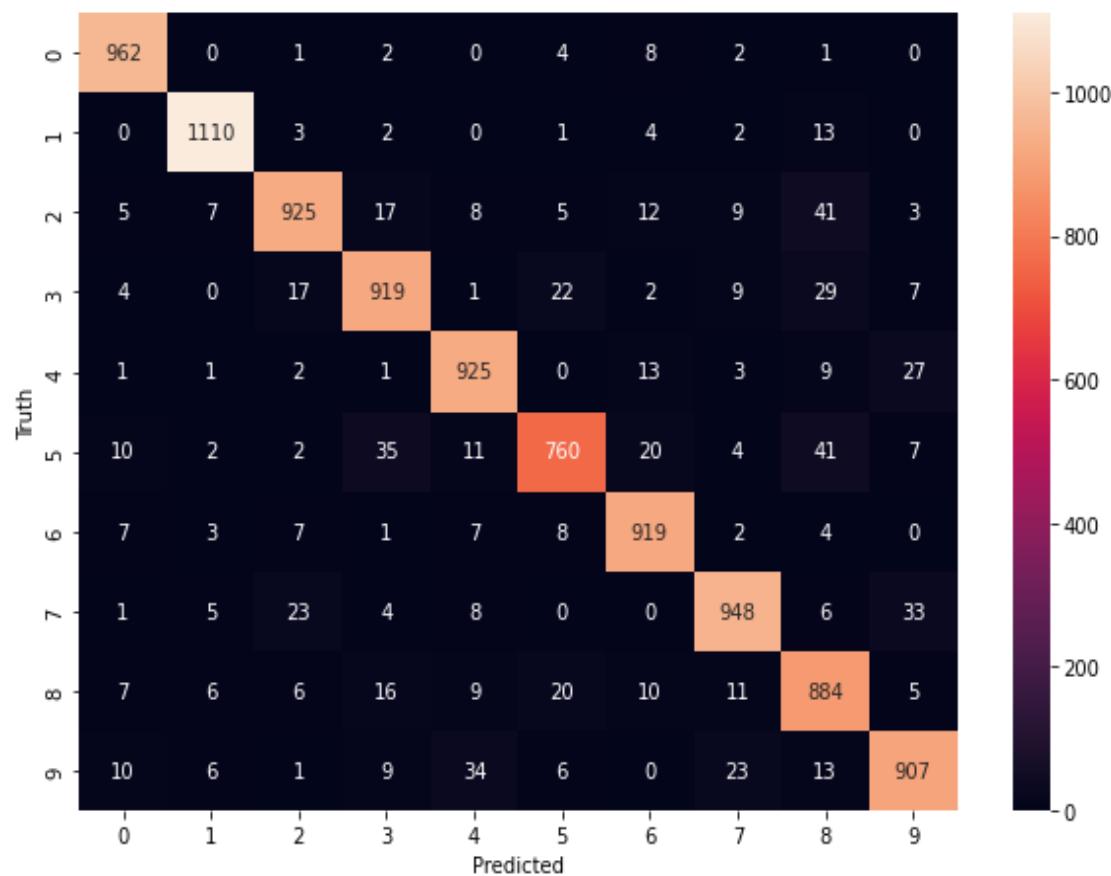
Neural Network

Accuracy = 0.9258999824523926

Classification Report

	precision	recall	f1-score	support
0	0.95	0.98	0.97	980
1	0.97	0.98	0.97	1135
2	0.93	0.90	0.91	1032
3	0.90	0.91	0.91	1010
4	0.93	0.93	0.93	982
5	0.90	0.87	0.89	892
6	0.94	0.95	0.95	958
7	0.93	0.93	0.93	1028
8	0.87	0.89	0.88	974
9	0.91	0.91	0.91	1009
accuracy			0.93	10000
macro avg	0.92	0.92	0.92	10000
weighted avg	0.93	0.93	0.93	10000

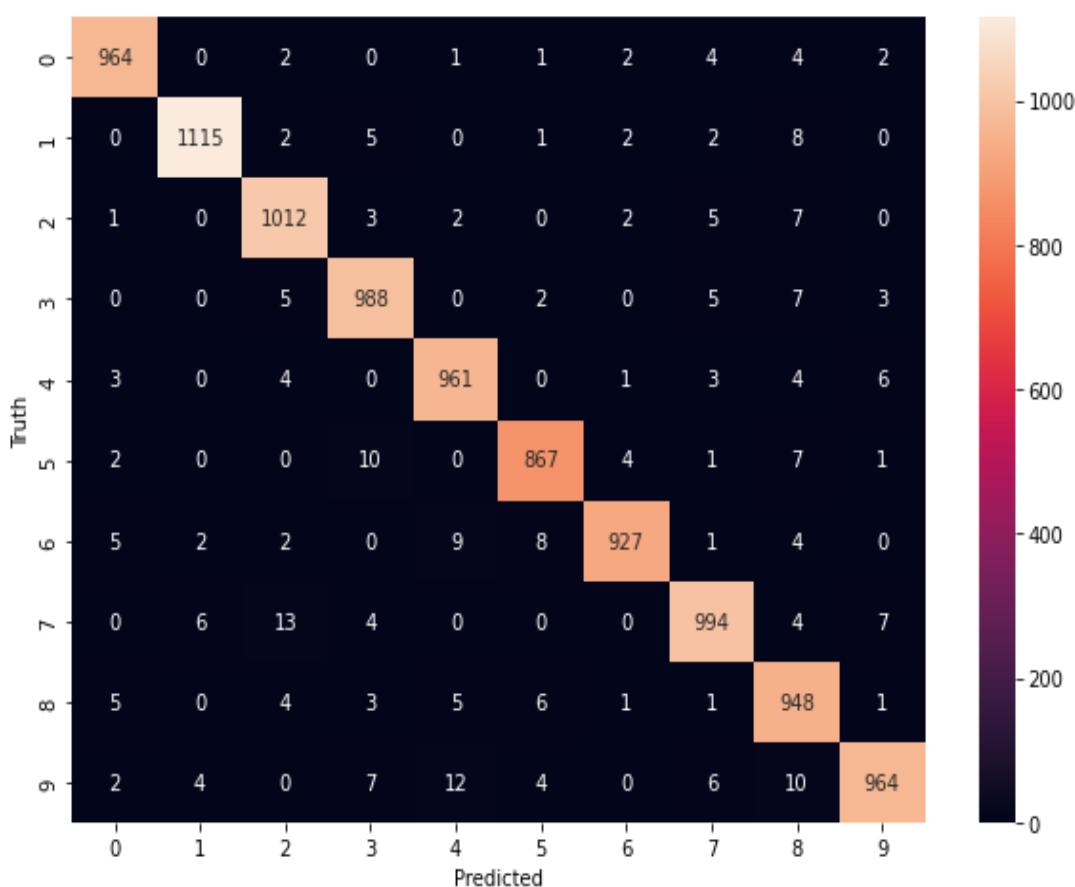
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Neural Network with a hidden layer	accuracy	score	=
0	0.96	0.98	0.97	980				
1	0.97	0.98	0.98	1135				
2	0.94	0.90	0.92	1032				
3	0.91	0.91	0.91	1010				
4	0.92	0.94	0.93	982				
5	0.92	0.85	0.88	892				
6	0.93	0.96	0.94	958				
7	0.94	0.92	0.93	1028				
8	0.85	0.91	0.88	974				
9	0.92	0.90	0.91	1009				
accuracy			0.93	10000				
macro avg	0.93	0.92	0.92	10000				
weighted avg	0.93	0.93	0.93	10000				

Confusion Matrix

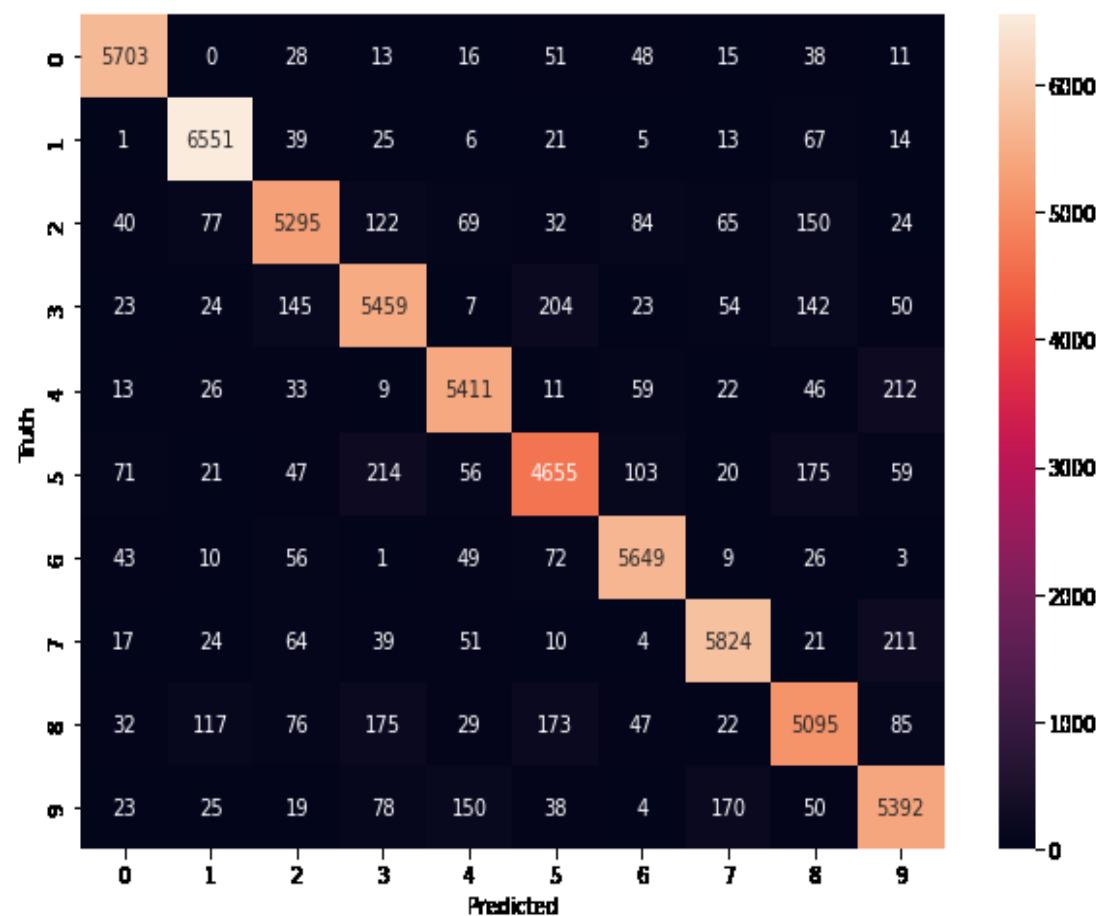


Classification Report

	precision	recall	f1-score	support	Regression Model	Cross-Validation Scores
0	0.98	0.98	0.98	980		
1	0.99	0.98	0.99	1135		
2	0.97	0.98	0.97	1032		
3	0.97	0.98	0.97	1010		
4	0.97	0.98	0.97	982		
5	0.98	0.97	0.97	892		
6	0.99	0.97	0.98	958		
7	0.97	0.97	0.97	1028		
8	0.95	0.97	0.96	974		
9	0.98	0.96	0.97	1009		
accuracy			0.97	10000		
macro avg	0.97	0.97	0.97	10000		
weighted avg	0.97	0.97	0.97	10000		

Accuracy = 0.9172333333333333

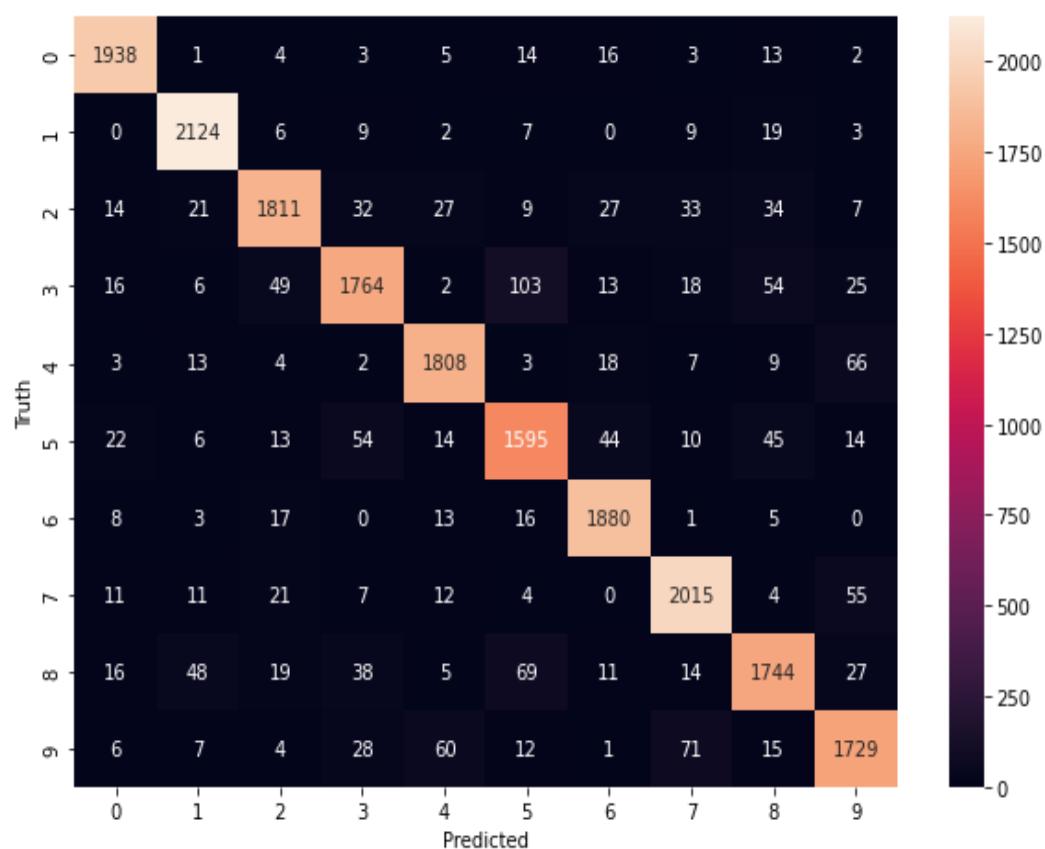
Confusion Matrix



	precision	recall	f1-score	support	Neural Network Scores	Validation Accuracy
0	0.96	0.96	0.96	5923		
1	0.95	0.97	0.96	6742		
2	0.91	0.89	0.90	5958		
3	0.89	0.89	0.89	6131		
4	0.93	0.93	0.93	5842		
5	0.88	0.86	0.87	5421		
6	0.94	0.95	0.95	5918		
7	0.94	0.93	0.93	6265		
8	0.88	0.87	0.87	5851		
9	0.89	0.91	0.90	5949		
accuracy			0.92	60000		
macro avg	0.92	0.92	0.92	60000		
weighted avg	0.92	0.92	0.92	60000		

Classification Report

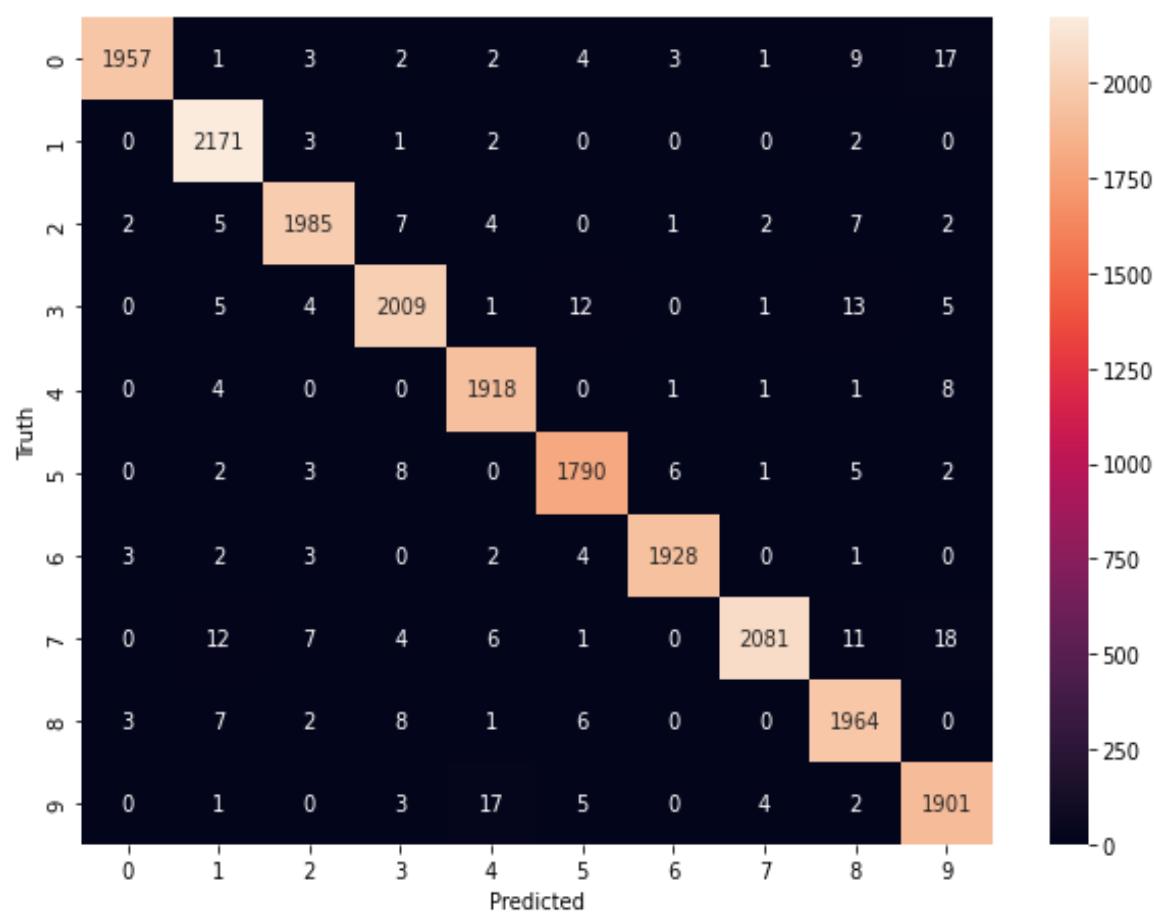
Confusion Matrix



	precision	recall	f1-score	support	
0	0.95	0.97	0.96	1999	
1	0.95	0.97	0.96	2179	
2	0.93	0.90	0.91	2015	
3	0.91	0.86	0.88	2050	Neural Network with a
4	0.93	0.94	0.93	1933	hidden layer
5	0.87	0.88	0.87	1817	Cross-Validation Scores
6	0.94	0.97	0.95	1943	Accuracy
7	0.92	0.94	0.93	2140	=
8	0.90	0.88	0.89	1991	0.9242333372433981
9	0.90	0.89	0.90	1933	
accuracy			0.92	20000	
macro avg	0.92	0.92	0.92	20000	
weighted avg	0.92	0.92	0.92	20000	

Classification Report

Confusion Matrix

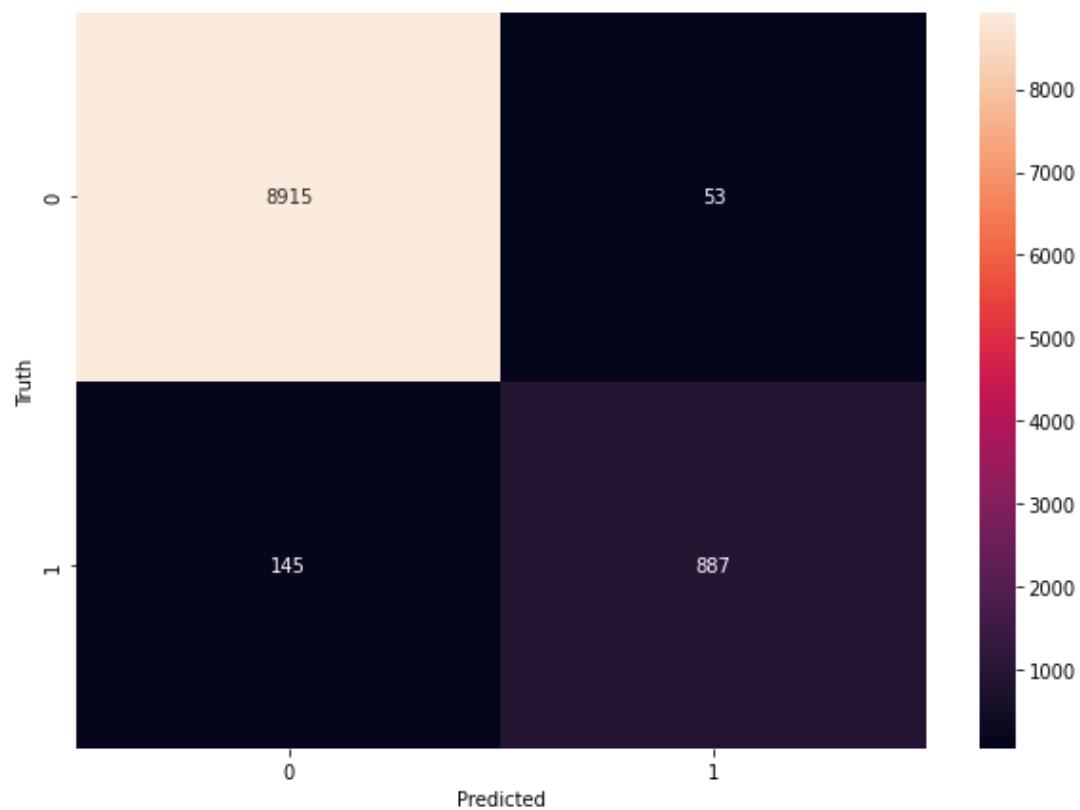


	precision	recall	f1-score	support
0	1.00	0.98	0.99	1999
1	0.98	1.00	0.99	2179
2	0.99	0.99	0.99	2015
3	0.98	0.98	0.98	2050
4	0.98	0.99	0.99	1933
5	0.98	0.99	0.98	1817
6	0.99	0.99	0.99	1943
7	1.00	0.97	0.98	2140
8	0.97	0.99	0.98	1991
9	0.97	0.98	0.98	1933
accuracy			0.99	20000
macro avg	0.99	0.99	0.99	20000
weighted avg	0.99	0.99	0.99	20000

B. Results for the
 '2' Detector
 Regression
 Model
 Accuracy =
 0.9802

Classification Report

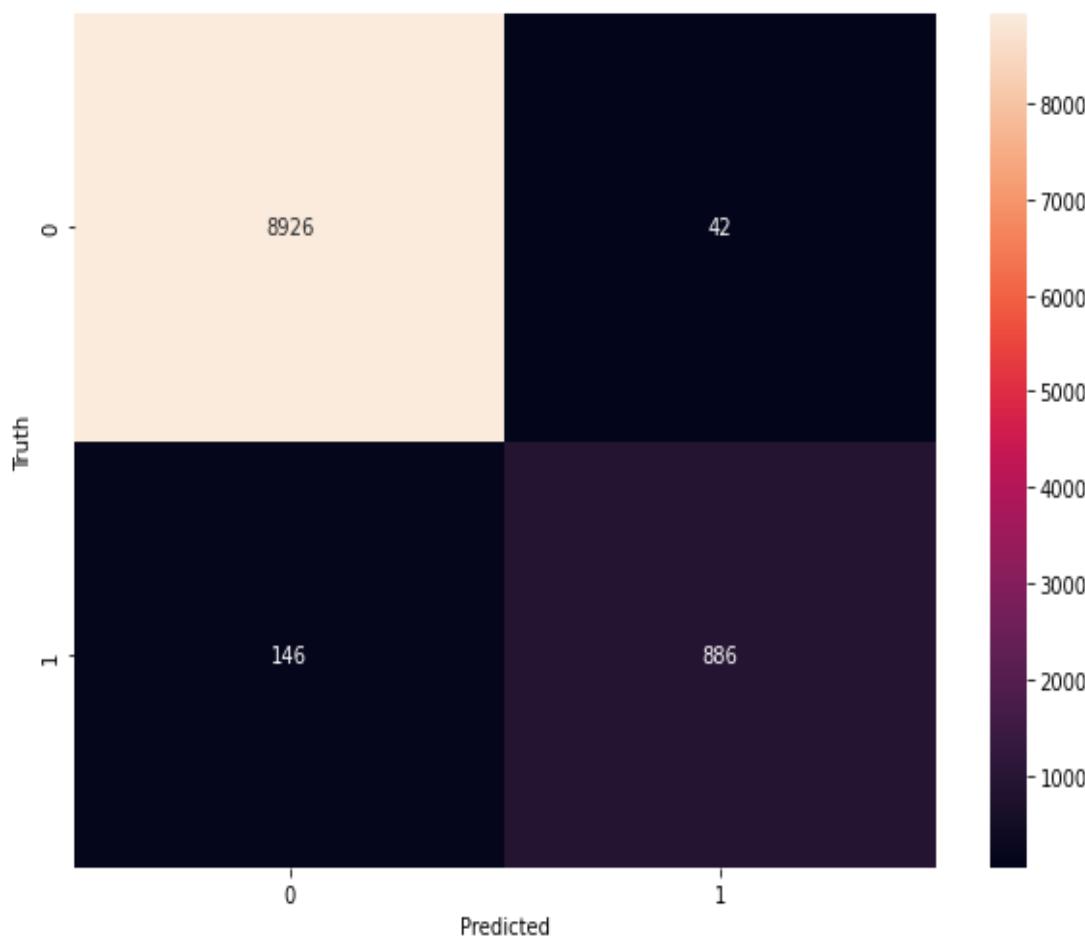
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Neural Network
False	0.98	0.99	0.99	8968	Accuracy =
True	0.94	0.86	0.90	1032	0.9811999797821045
accuracy			0.98	10000	
macro avg	0.96	0.93	0.94	10000	
weighted avg	0.98	0.98	0.98	10000	

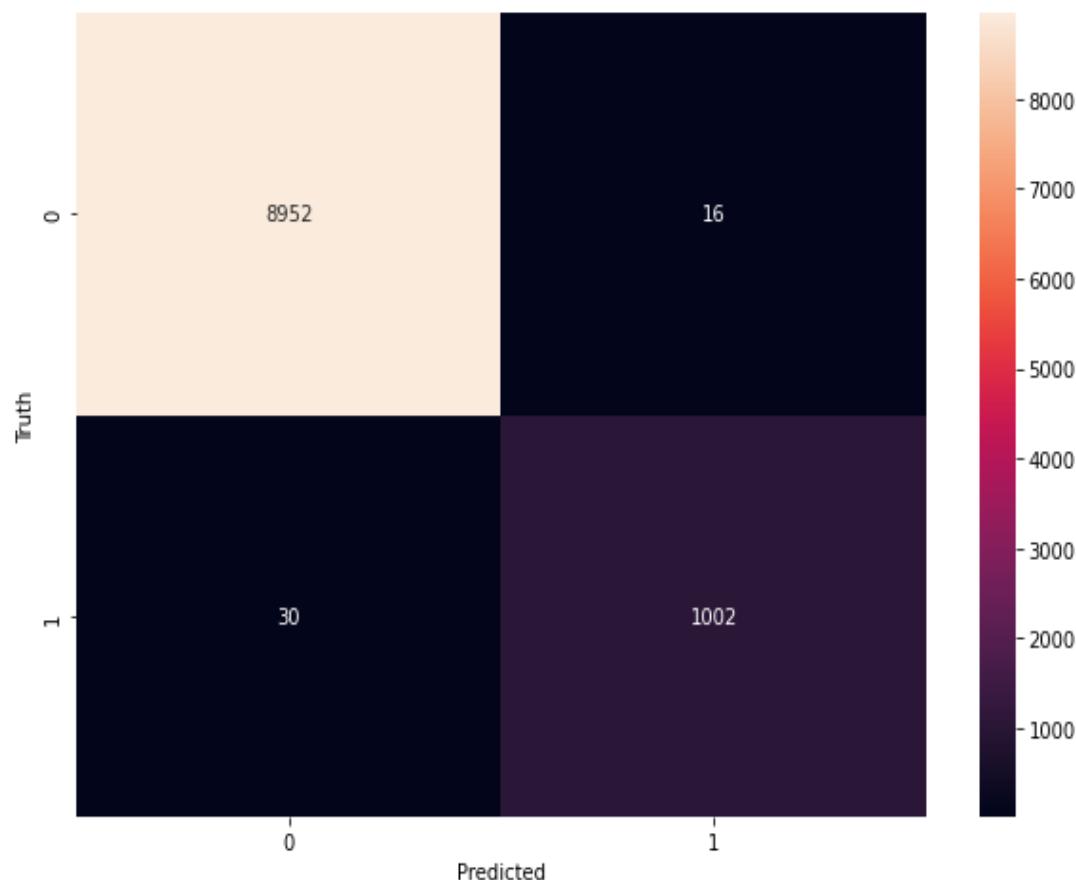
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Neural Network with a hidden layer
False	0.98	1.00	0.99	8968	Accuracy =
True	0.95	0.86	0.90	1032	
accuracy			0.98	10000	
macro avg	0.97	0.93	0.95	10000	
weighted avg	0.98	0.98	0.98	10000	
			0.9954000115394592		

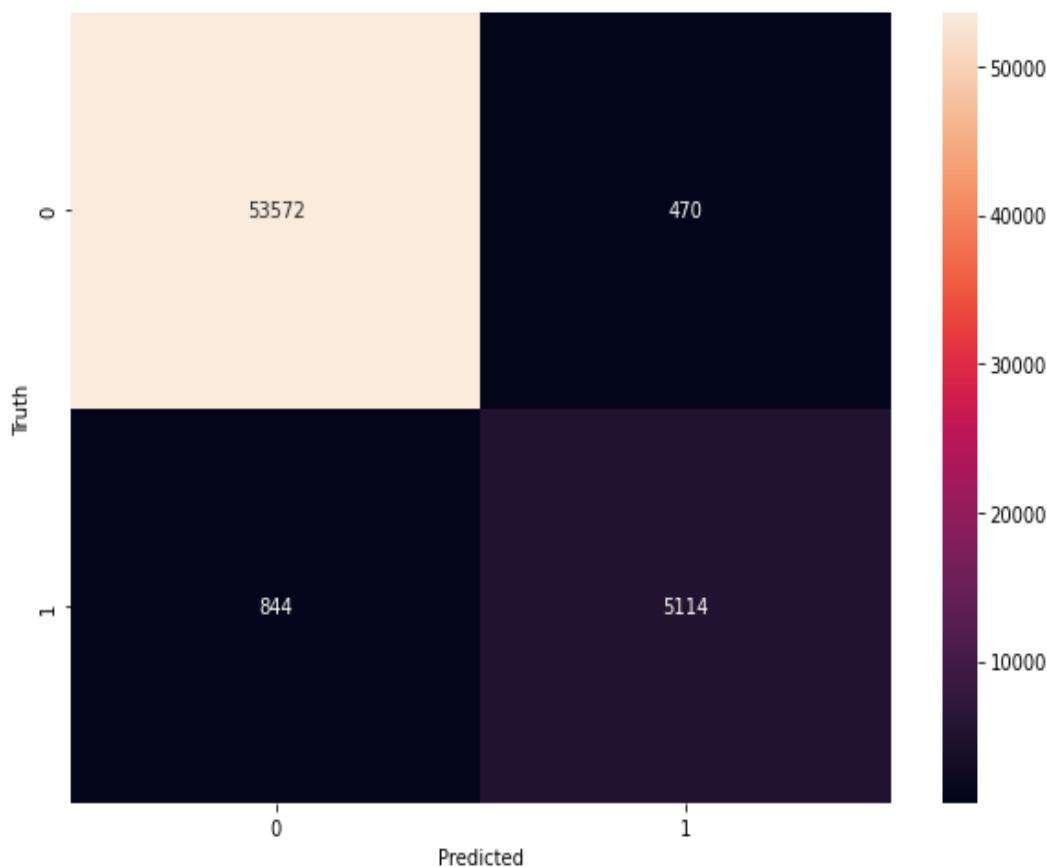
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Regression Model	Cross-Validation Scores
False	1.00	1.00	1.00	8968		
True	0.98	0.97	0.98	1032		
accuracy			1.00	10000		
macro avg	0.99	0.98	0.99	10000		
weighted avg	1.00	1.00	1.00	10000		
					Model Accuracy	= 0.9781

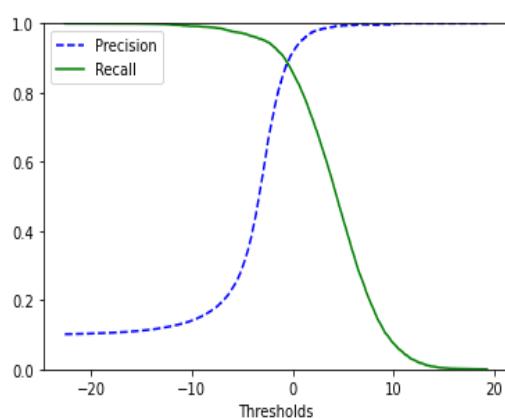
Confusion Matrix



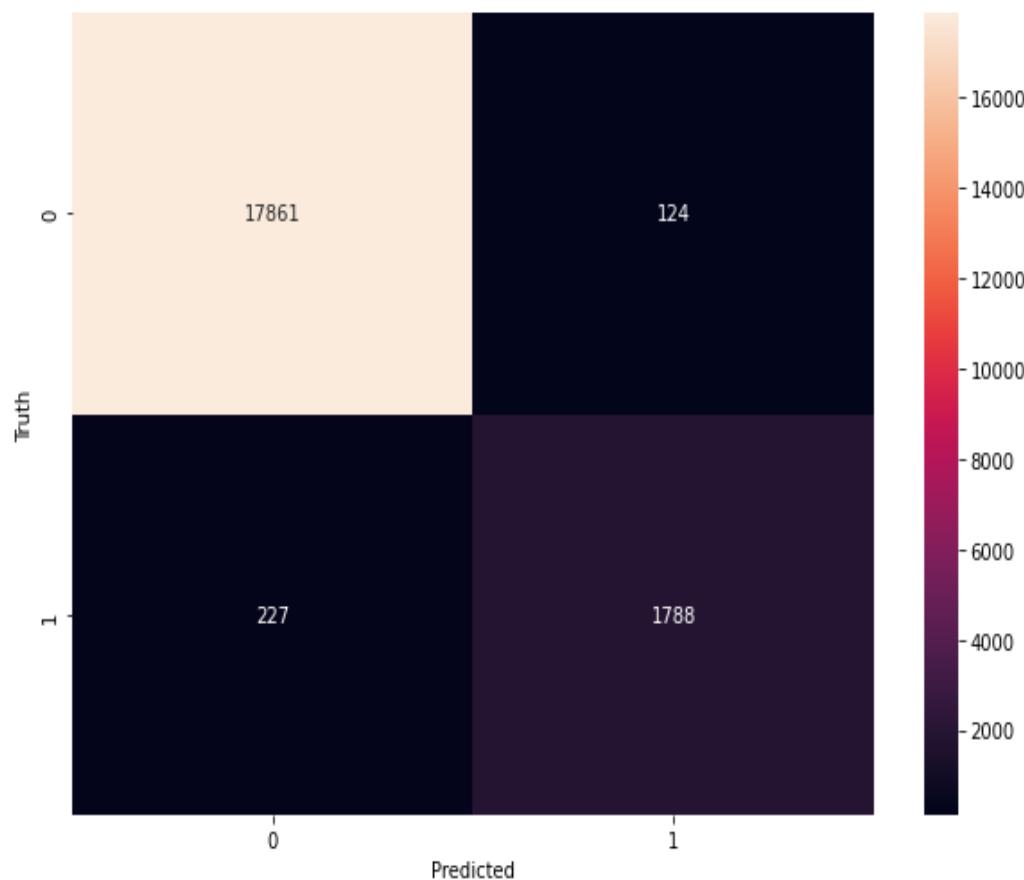
Classification Report

	precision	recall	f1-score	support	Neural Network Model Cross-Validation Scores
False	0.98	0.99	0.99	54042	Accuracy =
True	0.92	0.86	0.89	5958	
accuracy			0.98	60000	
macro avg	0.95	0.92	0.94	60000	
weighted avg	0.98	0.98	0.98	60000	
			0.9799000024795532		

Precision Recall Curve



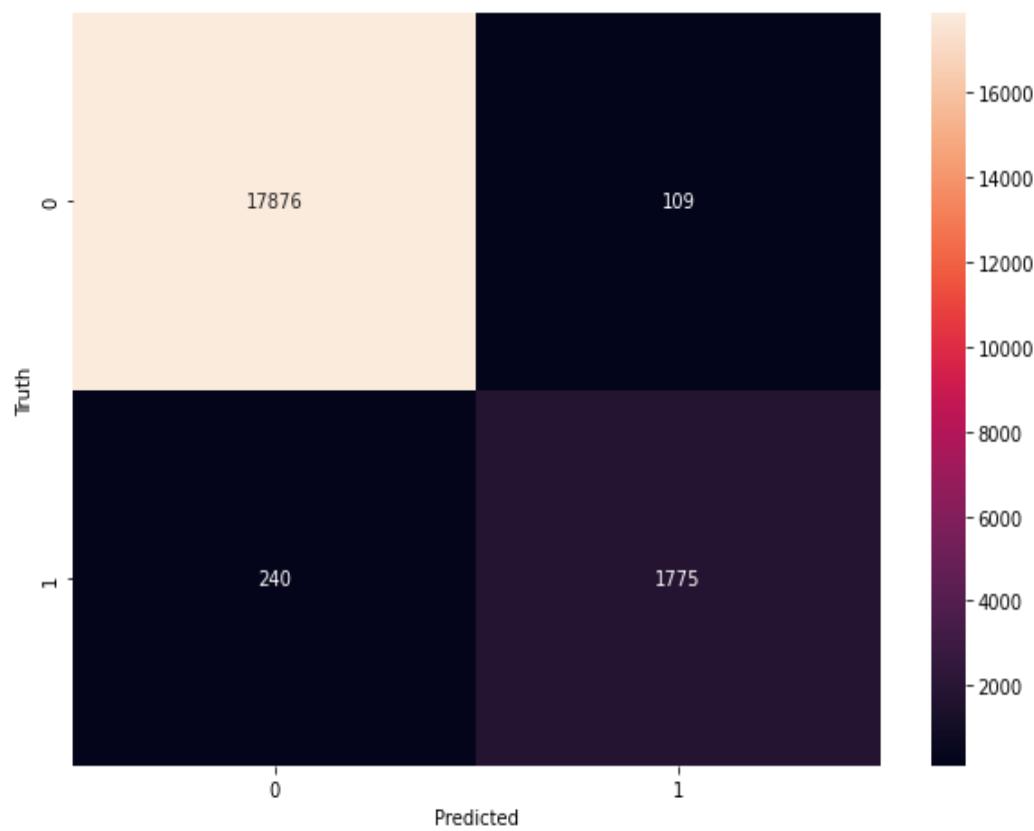
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Neural Network with a hidden layer Model Cross-Validation Scores
False	0.99	0.99	0.99	17985	Accuracy =
True	0.94	0.89	0.91	2015	
accuracy			0.98	20000	
macro avg	0.96	0.94	0.95	20000	
weighted avg	0.98	0.98	0.98	20000	
			0.9799166719118754		

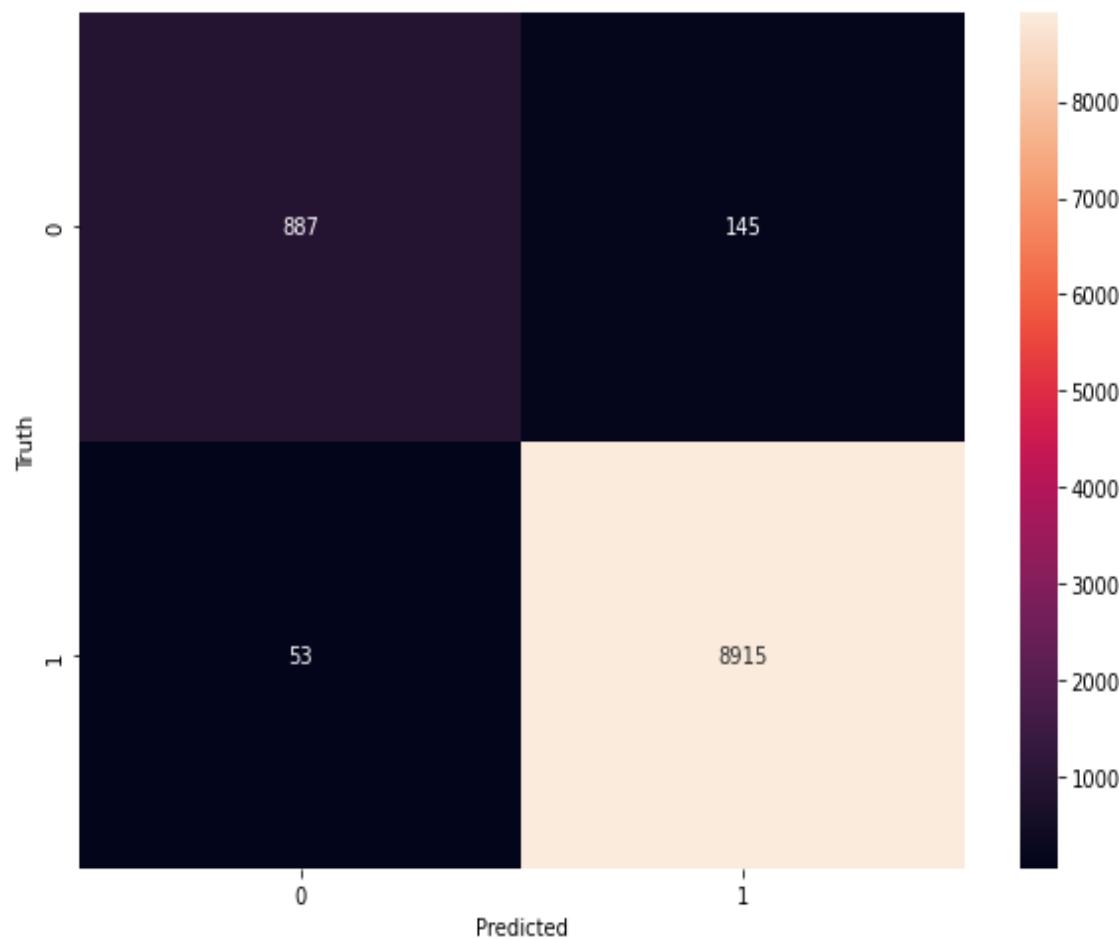
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	C. Results for the 'not 2' Detector Regression Model
False	0.99	0.99	0.99	17985	
True	0.94	0.88	0.91	2015	
accuracy			0.98	20000	Accuracy = 0.9802
macro avg	0.96	0.94	0.95	20000	
weighted avg	0.98	0.98	0.98	20000	

Confusion Matrix

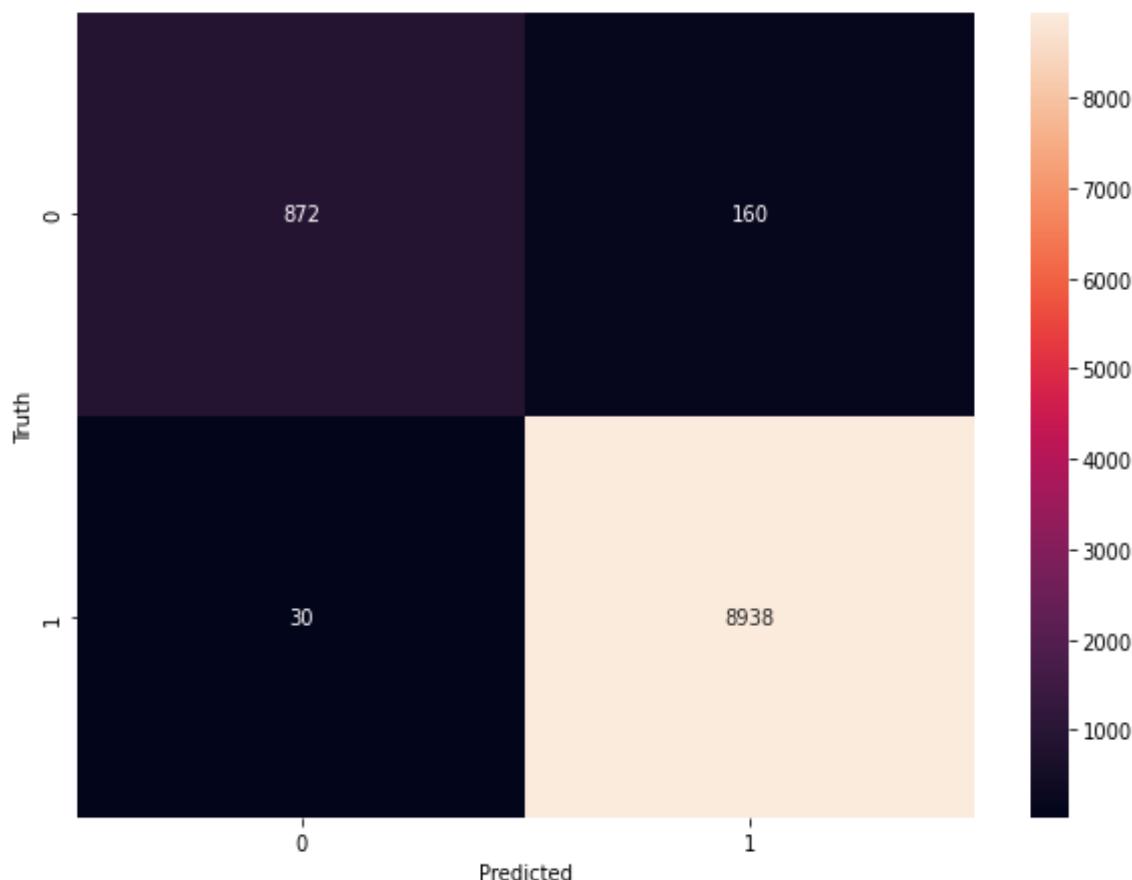


Classification Report

Neural Network Model
Accuracy = 0.9810000061988831

	precision	recall	f1-score	support
False	0.94	0.86	0.90	1032
True	0.98	0.99	0.99	8968
accuracy			0.98	10000
macro avg	0.96	0.93	0.94	10000
weighted avg	0.98	0.98	0.98	10000

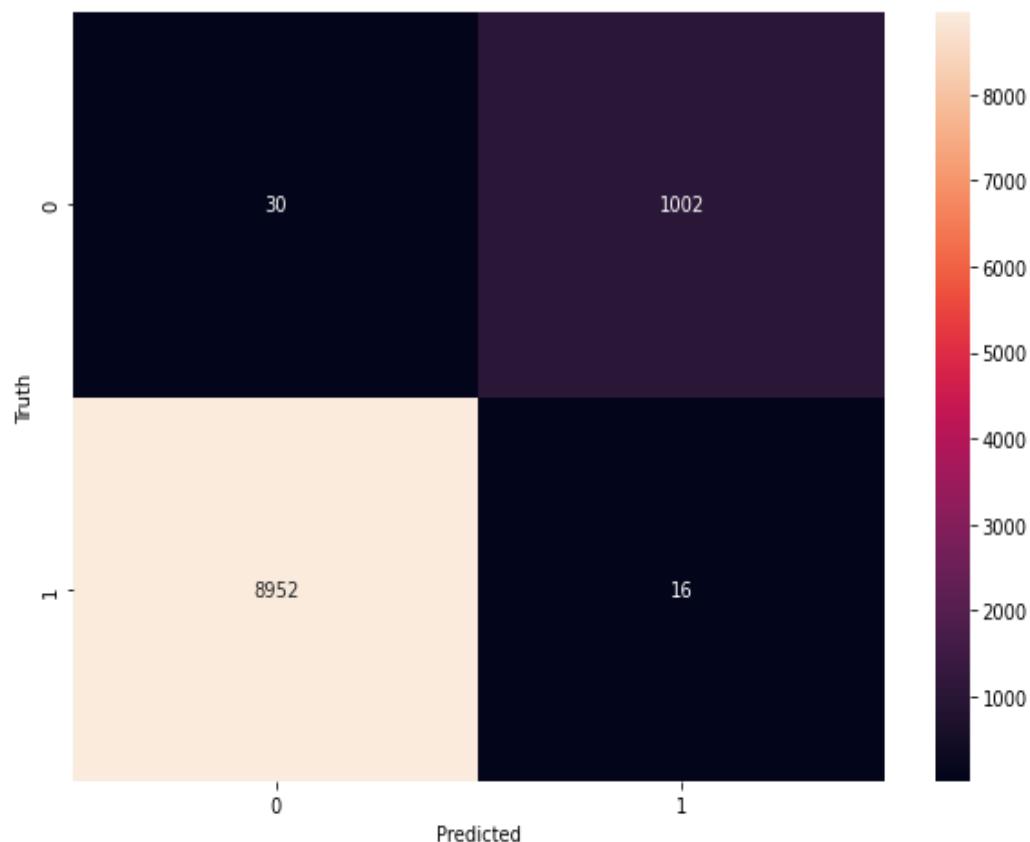
Confusion Matrix



Classification Report

	precision	recall	f1-score	support	Neural Network with a hidden layer Model
False	0.97	0.84	0.90	1032	
True	0.98	1.00	0.99	8968	Accuracy =
accuracy			0.98	10000	
macro avg	0.97	0.92	0.95	10000	
weighted avg	0.98	0.98	0.98	10000	
			0.9948999881744385		

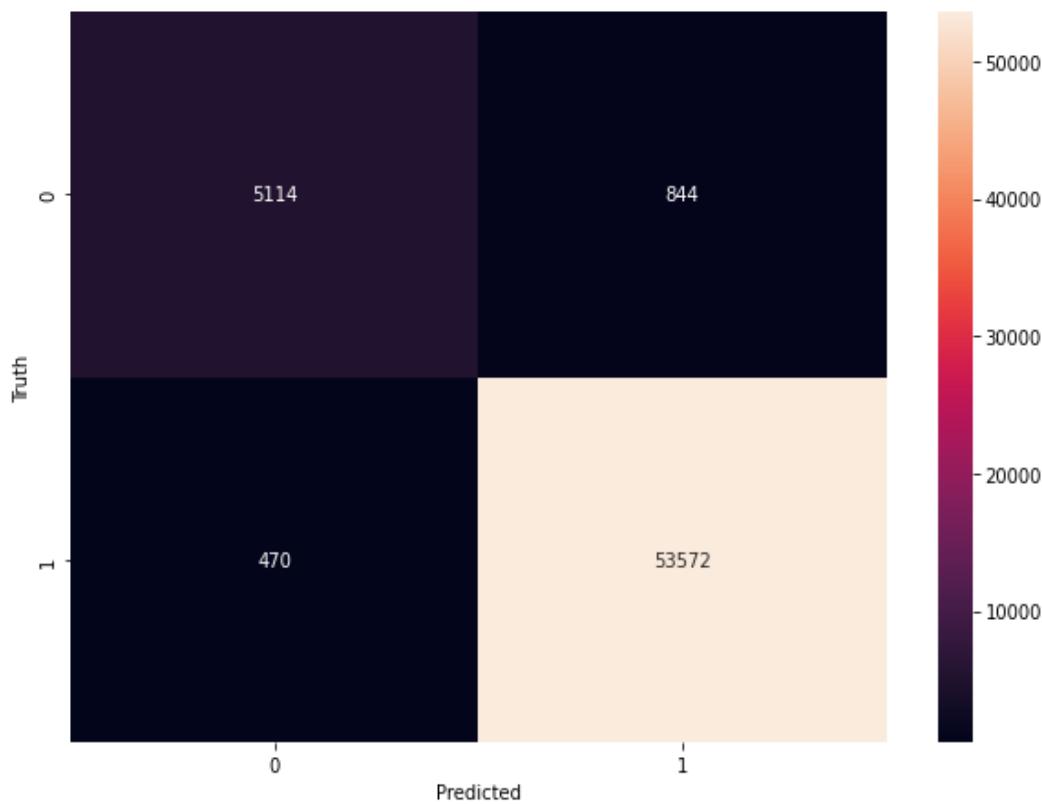
Confusion Matrix



Classification Report

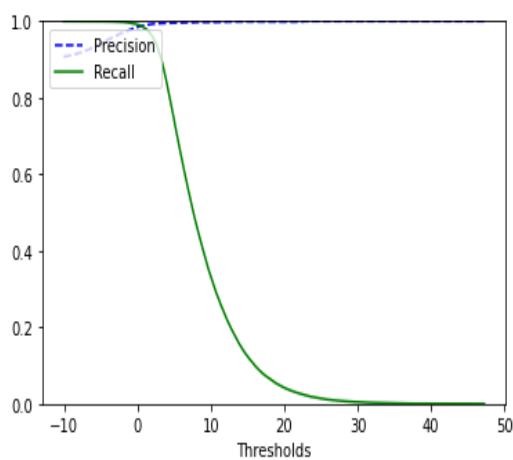
	precision	recall	f1-score	support	Regression	Model
False	0.98	0.97	0.98	1032	Cross Validation Scores Accuracy = 0.9781	
True	1.00	1.00	1.00	8968		
accuracy			0.99	10000		
macro avg	0.99	0.98	0.99	10000		
weighted avg	0.99	0.99	0.99	10000		

Confusion Matrix



Classification Report

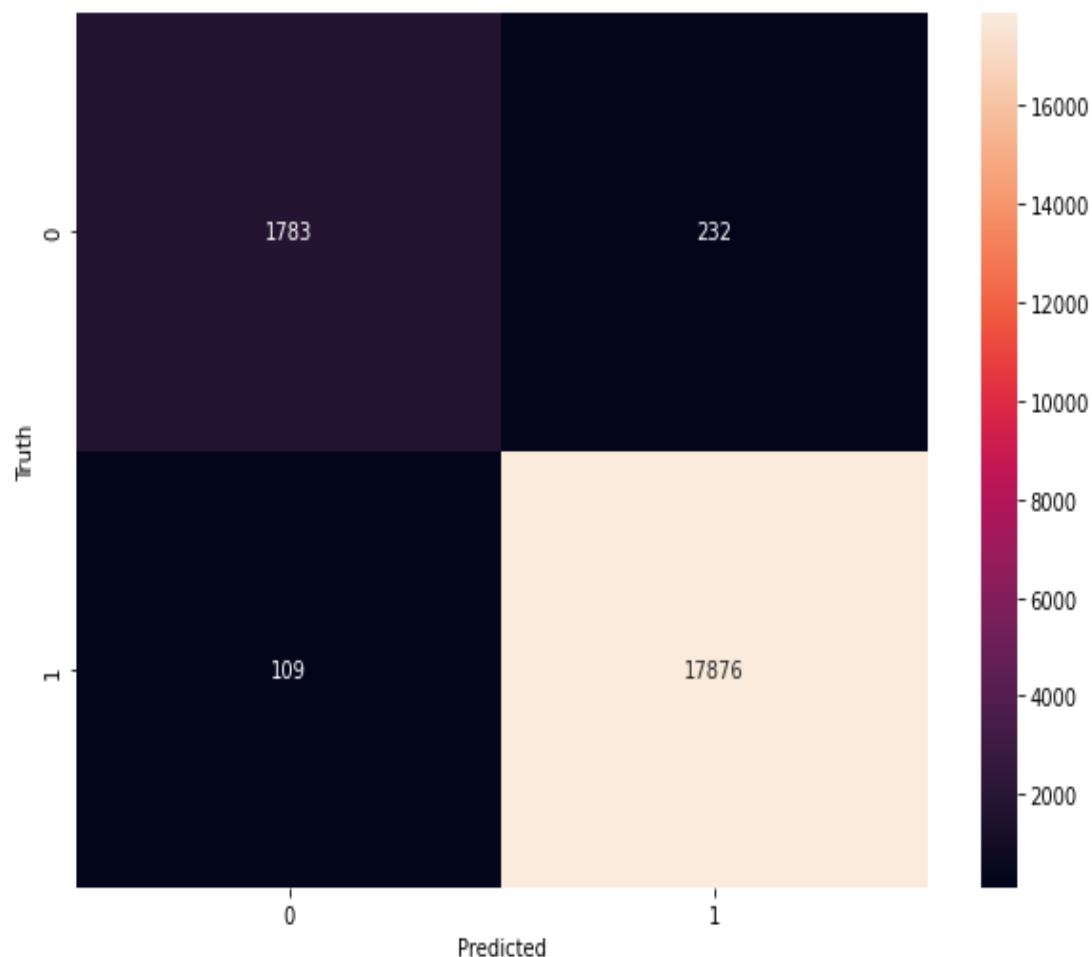
	precision	recall	f1-score	support
False	0.92	0.86	0.89	5958
True	0.98	0.99	0.99	54042
accuracy			0.98	60000
macro avg	0.95	0.92	0.94	60000
weighted avg	0.98	0.98	0.98	60000



Precision-Recall Curve

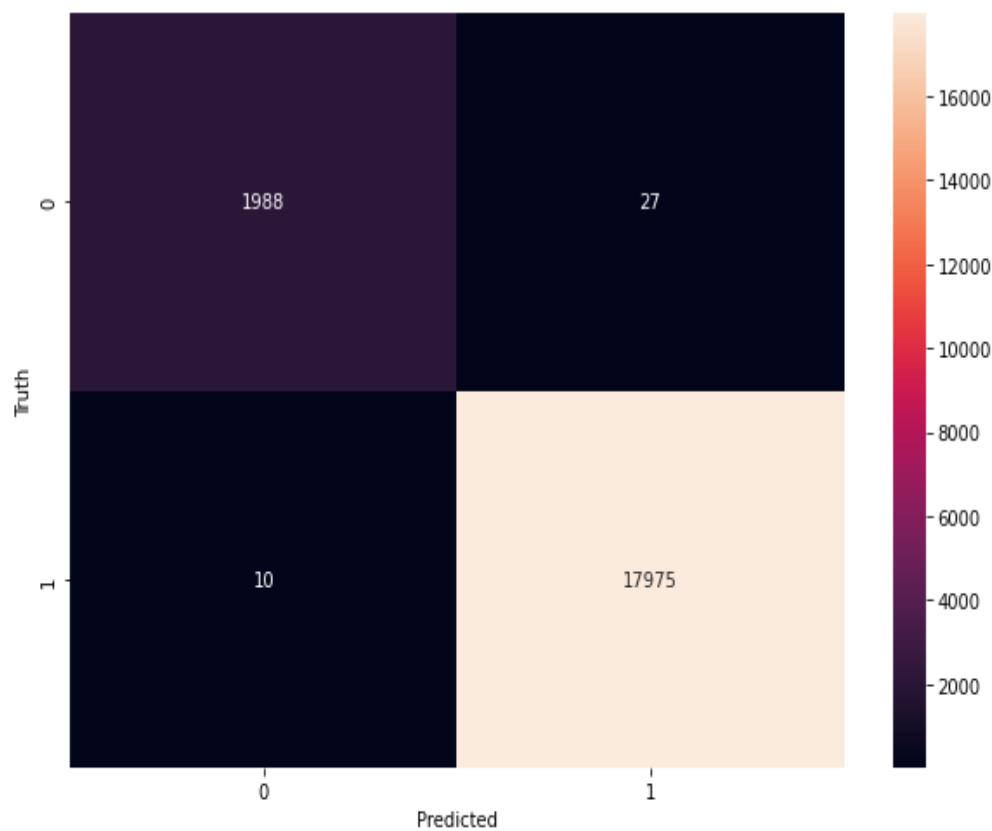
Neural Network Model Cross Validation Scores
Accuracy = 0.9802666505177816

Confusion Matrix



Classification Report		Network with a hidden layer		Model Scores	Neural Cross Validation	
		precision	recall	f1-score	support	Accuracy
False		0.94	0.88	0.91	2015	0.9976666569709778
True		0.99	0.99	0.99	17985	
accuracy				0.98	20000	
macro avg		0.96	0.94	0.95	20000	
weighted avg		0.98	0.98	0.98	20000	

Confusion Matrix



Classification Report

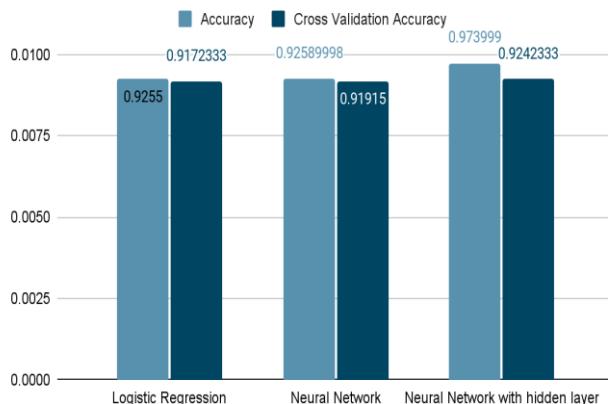
	precision	recall	f1-score	support
False	0.99	0.99	0.99	2015
True	1.00	1.00	1.00	17985
accuracy			1.00	20000
macro avg	1.00	0.99	0.99	20000
weighted avg	1.00	1.00	1.00	20000

IV. CONCLUSION

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A. For the full dataset classifiers

Full Dataset Classifier



Looking at the accuracy scores

The neural network is 0.0432177% better than the regression classifier

The Neural network with a hidden layer is 5.2403% better than the regression classifier

The neural network with a hidden layer is 5.19484% better than the neural network with no hidden layers

Now looking at the cross-validation accuracies

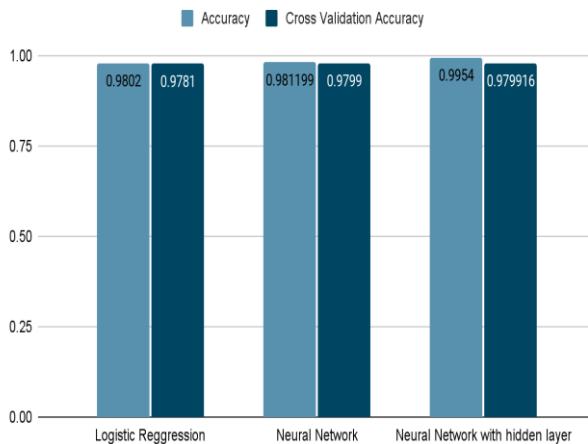
The neural network is 0.208965% better than the regression classifier

The Neural network with a hidden layer is 0.763165% better than the regression classifier

The neural network with a hidden layer is 0.553044% better than the neural network with no hidden layers

B. For the detect 'not 2' classifiers

'2' Classifier



Looking at the accuracy scores

The neural network is 0.101918% better than the regression classifier

The Neural network with a hidden layer is 1.5507% better than the regression classifier

The neural network with a hidden layer is 1.44731% better than the neural network with no hidden layers

Now looking at the cross-validation accuracies

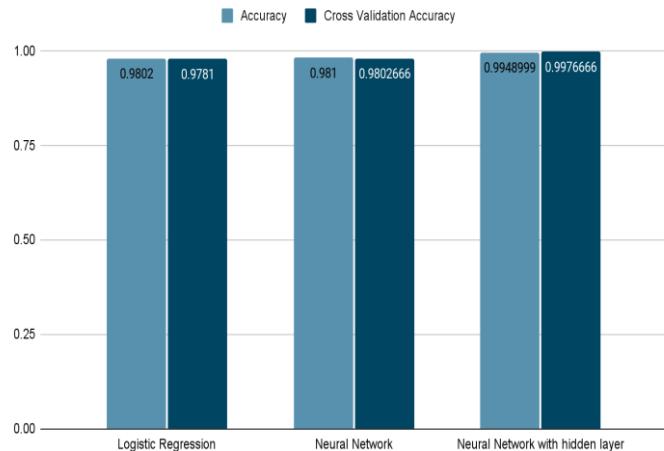
The neural network is 0.18403% better than the regression classifier

The Neural network with a hidden layer is 0.185666% better than the regression classifier

The neural network with a hidden layer is 0.00163282% better than the neural network with no hidden layers

C. For the detect 'not 2' classifiers

'Not 2 Classifier'



Looking at the accuracy scores

The neural network is 0.081616% better than the regression classifier

The Neural network with a hidden layer is 1.49968% better than the regression classifier

The neural network with a hidden layer is 1.41691% better than the neural network with no hidden layers

Now looking at the cross-validation accuracies

The neural network is 0.221511% better than the regression classifier

The Neural network with a hidden layer is 2.00047% better than the regression classifier

The neural network with a hidden layer is 1.77503% better than the neural network with no hidden layers

In each case, we can see that the Neural network with a single hidden layer seems to be better than the regression classifier while the neural network without any hidden layer is also better but only to a very small degree.