

# Hand Written Digits(MNIST) ML Project

-Pandey Shourya Prasad

## Overview of project:

I used the Handwritten digits dataset which had 60,000 training cases and 10,000 testing cases.

```
In [3]: (X_train, y_train), (X_test, y_test) = keras.datasets.mnist.load_data()

In [4]: len(X_train)
Out[4]: 60000

In [28]: len(X_test)
Out[28]: 10000
```

Tools used: Tensorflow, NumPy, matplotlib and Keras(in Tensorflow)

I basically made 5 models, each having a different activation function and losses function. Through this, I tried to figure out the model which had the highest accuracy during learning the training test.

I have used adaptive movement estimation(adam) in all the models.(Initial rate was set to be remain as default)

I have used the Dense layer in all the models.(Dense layer takes in use of all the neurons coming from previous layer/input)

Test Dataset had 60,000 numbers which basically had an array of the intensity of each block of 28x28 grid. So, I had to flatten that data set into  $(28*28,1) = (784,1)$  for the first layer.

## Model1 :

```
In [12]: model1=keras.Sequential([
          keras.layers.Dense(15,activation="sigmoid"),
          keras.layers.Dense(10,activation="sigmoid")
        ])
model1.compile(
    optimizer="adam",
    loss="sparse_categorical_crossentropy",
    metrics=['accuracy']
)
model1.fit(X_train_flattened,y_train, epochs=5)

Epoch 1/5
1875/1875 [=====] - 1s 721us/step - loss: 0.8030 - accuracy: 0.8315
Epoch 2/5
1875/1875 [=====] - 1s 681us/step - loss: 0.3459 - accuracy: 0.9086
Epoch 3/5
1875/1875 [=====] - 1s 684us/step - loss: 0.2790 - accuracy: 0.9228
Epoch 4/5
```

This model has two layers:-

- i) First layer has 15 units and activation function sigmoid
- ii) Second layer has 10 units and activation function sigmoid

The loss function used in this model is sparse categorical cross entropy

Dataset was trained 5 times(epochs=5) and at the end accuracy was 93.59%

## **Model2:**

```
In [13]: model2=keras.Sequential([
          keras.layers.Dense(15,activation="sigmoid"),
          keras.layers.Dense(10,activation="softmax")
        ])
model2.compile(
    optimizer="adam",
    loss="sparse_categorical_crossentropy",
    metrics=['accuracy']
)
model2.fit(X_train_flattened,y_train, epochs=5)

Epoch 1/5
1875/1875 [=====] - 1s 677us/step - loss: 0.8685 - accuracy: 0.8089
Epoch 2/5
1875/1875 [=====] - 1s 683us/step - loss: 0.3565 - accuracy: 0.9085
Epoch 3/5
1875/1875 [=====] - 1s 684us/step - loss: 0.2845 - accuracy: 0.9214
Epoch 4/5
1875/1875 [=====] - 1s 685us/step - loss: 0.2535 - accuracy: 0.9293
Epoch 5/5
1875/1875 [=====] - 1s 701us/step - loss: 0.2348 - accuracy: 0.9339

Out[13]: <keras.callbacks.History at 0x7fb93c7d0d60>
```

This model has two layers:-

- i) First layer has 15 units and activation function sigmoid
- ii) Second layer has 10 units and activation function softmax

The loss function used in this model is sparse categorical cross entropy.

Dataset was trained 5 times(epochs=5) and at the end accuracy was 93.39%

[ACCURACY NOT MUCH DIFFERENT FROM Model1]

## **Model3:**

```
In [18]: model3=keras.Sequential([
          keras.layers.Dense(15,activation="sigmoid"),
          keras.layers.Dense(12,activation="relu"),
          keras.layers.Dense(10,activation="linear")
        ])
model3.compile(
    optimizer="adam",
    loss="mean_squared_error",
    metrics=['accuracy']
)
model3.fit(X_train_flattened,y_train, epochs=5)

Epoch 1/5
1875/1875 [=====] - 2s 704us/step - loss: 3.9260 - accuracy: 0.0555
Epoch 2/5
1875/1875 [=====] - 2s 848us/step - loss: 1.2465 - accuracy: 0.0457
Epoch 3/5
1875/1875 [=====] - 1s 773us/step - loss: 1.0556 - accuracy: 0.0562
Epoch 4/5
1875/1875 [=====] - 1s 714us/step - loss: 0.9594 - accuracy: 0.0594
Epoch 5/5
1875/1875 [=====] - 1s 716us/step - loss: 0.8974 - accuracy: 0.0686

Out[18]: <keras.callbacks.History at 0x7fb93c2f4b80>
```

This model has three layers:-

- i) First layer has 15units and activation function sigmoid
- ii)Second layer has 12 units and activation function ReLU
- iii)Third layer has 10 units with activation function linear

The loss function used in this model is mean squared error.

Dataset was trained 5 times(epochs=5) and at the end accuracy was 06.86%

[ACCURACY WAS MUCH POORER THAN Model1 and Model2]

## Model4:

```
In [17]: model4=keras.Sequential([
          keras.layers.Dense(15,activation="sigmoid"),
          keras.layers.Dense(10,activation="softmax")
        ])
model4.compile(
    optimizer="adam",
    loss="mean_squared_error",
    metrics=['accuracy']
)
model4.fit(X_train_flattened,y_train, epochs=5)

Epoch 1/5
1875/1875 [=====] - 1s 688us/step - loss: 27.3046 - accuracy: 0.0992
Epoch 2/5
1875/1875 [=====] - 1s 702us/step - loss: 27.3046 - accuracy: 0.0932
Epoch 3/5
1875/1875 [=====] - 1s 702us/step - loss: 27.3046 - accuracy: 0.0930
Epoch 4/5
1875/1875 [=====] - 1s 695us/step - loss: 27.3046 - accuracy: 0.0941
Epoch 5/5
1875/1875 [=====] - 1s 782us/step - loss: 27.3046 - accuracy: 0.0989

Out[17]: <keras.callbacks.History at 0x7fb93c423a60>
```

This model has two layers:-

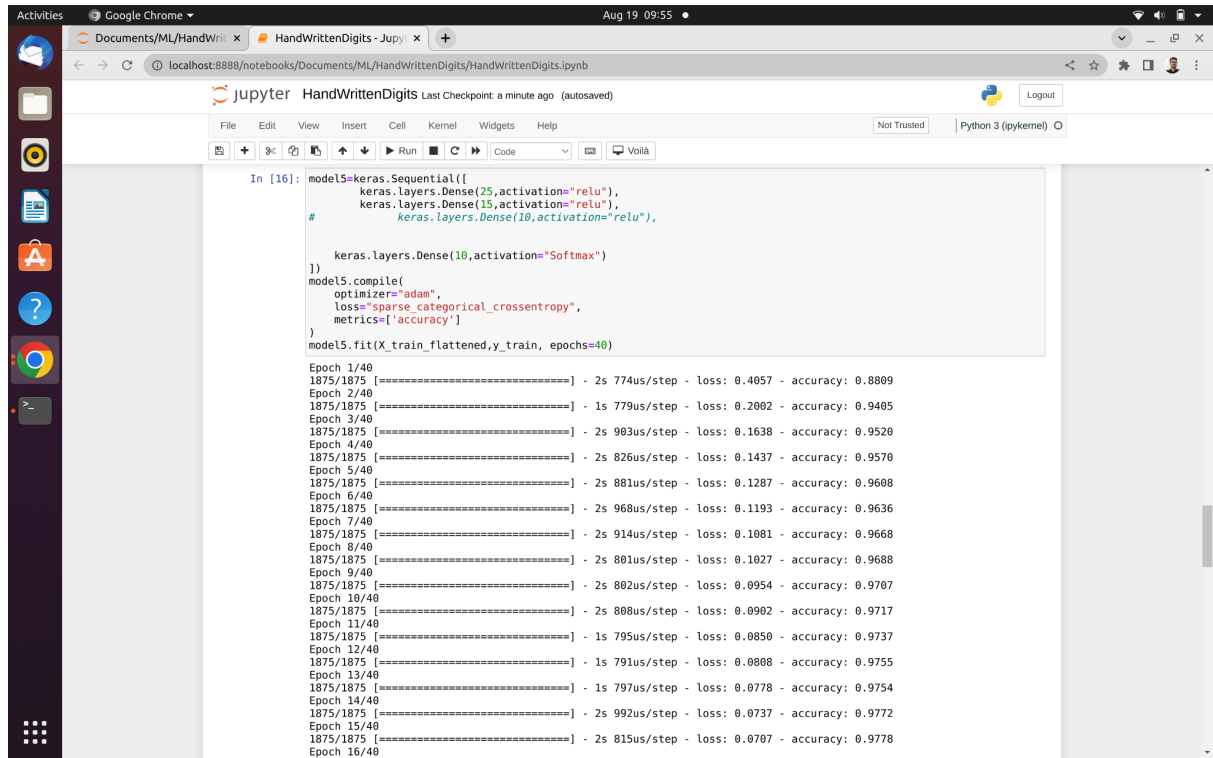
- i) First layer has 15units and activation function sigmoid
- ii)Second layer has 12 units and activation function softmax

The loss function used in this model is mean squared error.

Dataset was trained 5 times(epochs=5) and at the end accuracy was 09.89%.

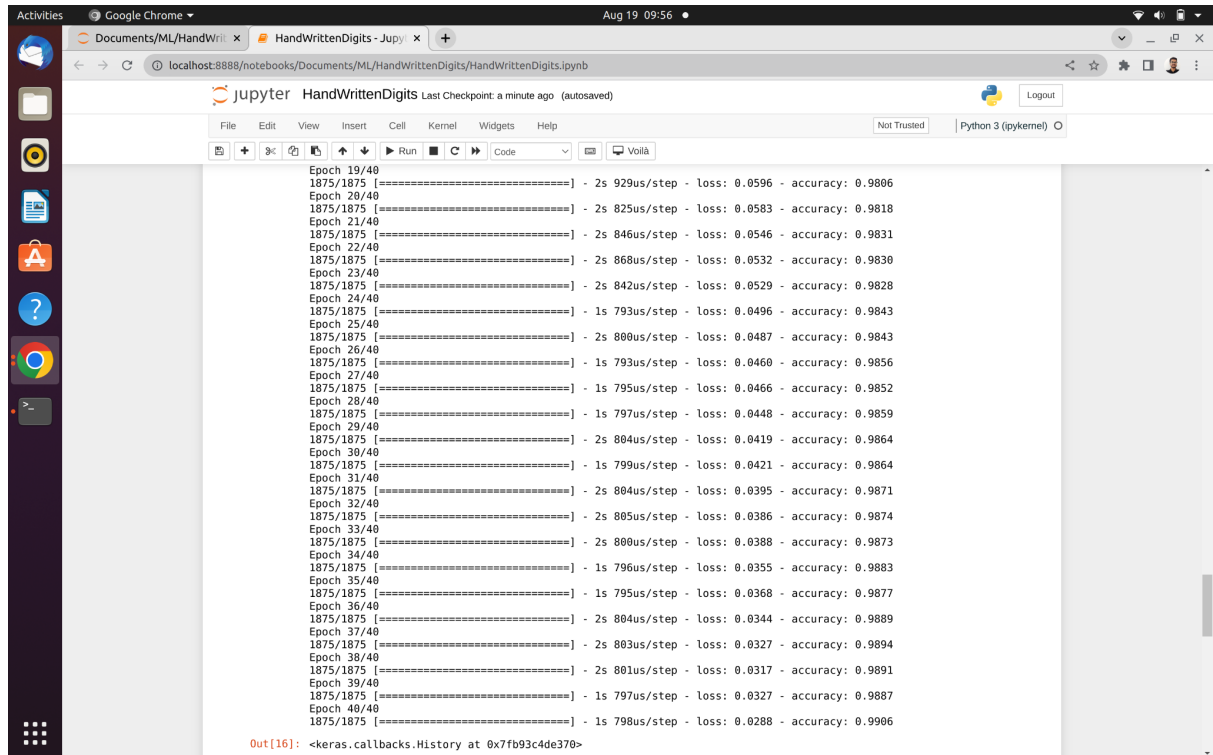
[ ACCURACY WAS MUCH POORER THAN Model1 and Model2(slightly higher than Model3) ]

# Model5:



```
In [16]: model5=keras.Sequential([
keras.layers.Dense(25,activation="relu"),
keras.layers.Dense(15,activation="relu"),
#
keras.layers.Dense(10,activation="relu"),
])
keras.layers.Dense(10,activation="Softmax")
model5.compile(
optimizer="adam",
loss="sparse_categorical_crossentropy",
metrics=['accuracy']
)
model5.fit(X_train_flattened,y_train, epochs=40)

Epoch 1/40
1875/1875 [=====] - 2s 774us/step - loss: 0.4057 - accuracy: 0.8809
Epoch 2/40
1875/1875 [=====] - 1s 779us/step - loss: 0.2002 - accuracy: 0.9405
Epoch 3/40
1875/1875 [=====] - 2s 903us/step - loss: 0.1638 - accuracy: 0.9520
Epoch 4/40
1875/1875 [=====] - 2s 826us/step - loss: 0.1437 - accuracy: 0.9570
Epoch 5/40
1875/1875 [=====] - 2s 881us/step - loss: 0.1287 - accuracy: 0.9608
Epoch 6/40
1875/1875 [=====] - 2s 968us/step - loss: 0.1193 - accuracy: 0.9636
Epoch 7/40
1875/1875 [=====] - 2s 914us/step - loss: 0.1081 - accuracy: 0.9668
Epoch 8/40
1875/1875 [=====] - 2s 801us/step - loss: 0.1027 - accuracy: 0.9688
Epoch 9/40
1875/1875 [=====] - 2s 802us/step - loss: 0.0954 - accuracy: 0.9707
Epoch 10/40
1875/1875 [=====] - 2s 808us/step - loss: 0.0902 - accuracy: 0.9717
Epoch 11/40
1875/1875 [=====] - 1s 795us/step - loss: 0.0850 - accuracy: 0.9737
Epoch 12/40
1875/1875 [=====] - 1s 791us/step - loss: 0.0808 - accuracy: 0.9755
Epoch 13/40
1875/1875 [=====] - 1s 797us/step - loss: 0.0778 - accuracy: 0.9754
Epoch 14/40
1875/1875 [=====] - 2s 992us/step - loss: 0.0737 - accuracy: 0.9772
Epoch 15/40
1875/1875 [=====] - 2s 815us/step - loss: 0.0707 - accuracy: 0.9778
Epoch 16/40
```



```
Epoch 17/40
1875/1875 [=====] - 2s 929us/step - loss: 0.0596 - accuracy: 0.9806
Epoch 18/40
1875/1875 [=====] - 2s 825us/step - loss: 0.0583 - accuracy: 0.9818
Epoch 19/40
1875/1875 [=====] - 2s 846us/step - loss: 0.0546 - accuracy: 0.9831
Epoch 20/40
1875/1875 [=====] - 2s 868us/step - loss: 0.0532 - accuracy: 0.9830
Epoch 21/40
1875/1875 [=====] - 2s 842us/step - loss: 0.0529 - accuracy: 0.9828
Epoch 22/40
1875/1875 [=====] - 1s 793us/step - loss: 0.0496 - accuracy: 0.9843
Epoch 23/40
1875/1875 [=====] - 2s 800us/step - loss: 0.0487 - accuracy: 0.9843
Epoch 24/40
1875/1875 [=====] - 1s 793us/step - loss: 0.0460 - accuracy: 0.9856
Epoch 25/40
1875/1875 [=====] - 1s 795us/step - loss: 0.0466 - accuracy: 0.9852
Epoch 26/40
1875/1875 [=====] - 1s 797us/step - loss: 0.0448 - accuracy: 0.9859
Epoch 27/40
1875/1875 [=====] - 2s 804us/step - loss: 0.0419 - accuracy: 0.9864
Epoch 28/40
1875/1875 [=====] - 1s 799us/step - loss: 0.0421 - accuracy: 0.9864
Epoch 29/40
1875/1875 [=====] - 2s 804us/step - loss: 0.0395 - accuracy: 0.9871
Epoch 30/40
1875/1875 [=====] - 2s 805us/step - loss: 0.0386 - accuracy: 0.9874
Epoch 31/40
1875/1875 [=====] - 2s 800us/step - loss: 0.0388 - accuracy: 0.9873
Epoch 32/40
1875/1875 [=====] - 1s 796us/step - loss: 0.0355 - accuracy: 0.9883
Epoch 33/40
1875/1875 [=====] - 1s 795us/step - loss: 0.0368 - accuracy: 0.9877
Epoch 34/40
1875/1875 [=====] - 2s 804us/step - loss: 0.0344 - accuracy: 0.9889
Epoch 35/40
1875/1875 [=====] - 2s 803us/step - loss: 0.0327 - accuracy: 0.9894
Epoch 36/40
1875/1875 [=====] - 2s 801us/step - loss: 0.0317 - accuracy: 0.9891
Epoch 37/40
1875/1875 [=====] - 1s 797us/step - loss: 0.0327 - accuracy: 0.9887
Epoch 38/40
1875/1875 [=====] - 1s 798us/step - loss: 0.0288 - accuracy: 0.9906

Out[16]: <keras.callbacks.History at 0x7fb93c4de370>
```

This model has three layers:-

- i) First layer has 25 units and activation function ReLU
- ii) Second layer has 15 units and activation function ReLU
- iii) Third layer has 10 units with activation function Softmax

The loss function used in this model is sparse categorical cross entropy

Dataset was trained 40 times(epochs=40) and at the end accuracy was 99.06%(As it was giving the higher accuracy for 5 epochs so I tried making it learn 40 times so as to increase its accuracy)

[THIS MODEL WAS THE BEST MODEL AND HAD HIGHEST PERCENTAGE OF ACCURACY AMONG ALL THE FIVE MODELS]

## **OBSERVATION:**

I tried random examples and then predicted answer from all the models.

Eg:-



Testcase was 2306 and plt.matshow() helped me in showing the image of testcase

`y_test()` returns the answer associated with `X_test()`

```
In [21]: a=model1.predict(X_test_flattened)
         np.argmax(a[2306])
313/313 [=====] - 0s 424us/step
Out[21]: 4

In [22]: a=model2.predict(X_test_flattened)
         np.argmax(a[2306])
313/313 [=====] - 0s 455us/step
Out[22]: 3

In [23]: a=model3.predict(X_test_flattened)
         np.argmax(a[2306])
313/313 [=====] - 0s 464us/step
Out[23]: 4

In [24]: a=model4.predict(X_test_flattened)
         np.argmax(a[2306])
313/313 [=====] - 0s 426us/step
Out[24]: 6

In [25]: a=model5.predict(X_test_flattened)
         np.argmax(a[2306])
313/313 [=====] - 0s 469us/step
Out[25]: 4
```

We can notice that correct answer was predicted by model1, model3 and model5

Even though model2 had large accuracy but it predicted wrong answer and model3 had low accuracy but it predicted the correct answer.

`np.argmax()` returned the index of maximum value in the whole array.