

## Sprint - 4

Team ID	PNT2022TMID18280
Project Name	A Novel Method for Handwritten Digit Recognition

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## Importing Packages

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from keras.utils import np_utils
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, Dense, Flatten
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import load_model
from PIL import Image, ImageOps
import numpy
from keras.datasets import mnist
from matplotlib import pyplot
```

## Load the data

```
In [2]: (X_train, y_train), (X_test, y_test) = mnist.load_data()
```

### Analyse the data

```
In [3]: print(X_train.shape)
        print(X_test.shape)
```

(60000, 28, 28)  
(10000, 28, 28)

```
In [5]: X_train[2]
```

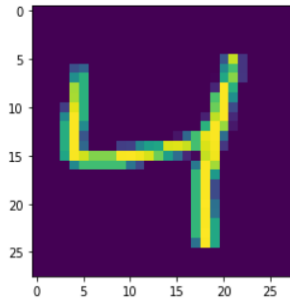
```
Out[5]: array([[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 67, 232, 39, 0, 0, 0, 0],
 [0, 0, 0, 0, 62, 81, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 0, 120, 180, 39, 0, 0, 0],
 [0, 0, 0, 0, 126, 163, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 2, 153, 210, 40, 0, 0, 0],
 [0, 0, 0, 0, 220, 163, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 183, 254, 125, 0, 0, 0, 0],
 [0, 0, 0, 46, 245, 163, 0, 0, 0, 0, 0, 0, 0],
 [0, 0, 0, 0, 0, 0, 198, 254, 56, 0, 0, 0, 0],
```

```
In [6]: y_train[2]
```

```
Out[6]: 4
```

```
In [7]: plt.imshow(X_train[2])
```

```
Out[7]: <matplotlib.image.AxesImage at 0x28248079e50>
```



#### Data Preprocessing

```
In [8]: X_train = X_train.reshape(60000, 28, 28, 1).astype('float32')
X_test = X_test.reshape(10000, 28, 28, 1).astype('float32')
```

```
In [9]: number_of_classes = 10
Y_train = np_utils.to_categorical(y_train, number_of_classes)
Y_test = np_utils.to_categorical(y_test, number_of_classes)
```

```
In [10]: Y_train[2]
```

```
Out[10]: array([0., 0., 0., 0., 1., 0., 0., 0., 0., 0.], dtype=float32)
```

#### Create the model

```
In [11]: model = Sequential()
model.add(Conv2D(64, (3, 3), input_shape=(28, 28, 1), activation="relu"))
model.add(Conv2D(32, (3, 3), activation="relu"))
model.add(Flatten())
model.add(Dense(number_of_classes, activation="softmax"))
```

#### Compile the model

```
In [12]: model.compile(loss='categorical_crossentropy', optimizer="Adam", metrics=["accuracy"])
```

#### Train the model

```
In [13]: model.fit(X_train, Y_train, batch_size=32, epochs=5, validation_data=(X_test, Y_test))
```

```
Epoch 1/5
1875/1875 [=====] - 109s 58ms/step - loss: 0.2591 - accuracy: 0.9469 - val_loss: 0.1248 - val_accuracy: 0.9634
Epoch 2/5
1875/1875 [=====] - 110s 58ms/step - loss: 0.0742 - accuracy: 0.9776 - val_loss: 0.0905 - val_accuracy: 0.9710
Epoch 3/5
1875/1875 [=====] - 109s 58ms/step - loss: 0.0474 - accuracy: 0.9845 - val_loss: 0.0855 - val_accuracy: 0.9758
Epoch 4/5
1875/1875 [=====] - 108s 58ms/step - loss: 0.0354 - accuracy: 0.9889 - val_loss: 0.1053 - val_accuracy: 0.9752
Epoch 5/5
1875/1875 [=====] - 106s 57ms/step - loss: 0.0281 - accuracy: 0.9915 - val_loss: 0.1086 - val_accuracy: 0.9748
```

```
Out[13]: <keras.callbacks.History at 0x282475e0250>
```

Test the model

```
In [14]: metrics = model.evaluate(X_test, Y_test, verbose=0)
print("Metrics (Test Loss & Test Accuracy): ")
print(metrics)
```

```
Metrics (Test Loss & Test Accuracy):
[0.1085574524168015, 0.9747999906539917]
```

```
In [15]: prediction = model.predict(X_test[:6])
print(prediction)
```

```
1/1 [=====] - 0s 171ms/step
[[1.2065910e-12 2.9789200e-14 1.2585115e-10 3.1951169e-10 1.6641216e-13
 5.5336404e-15 1.4522152e-20 1.0000000e+00 6.7726891e-09 6.7247728e-11]
 [1.5630913e-10 6.5942291e-10 1.0000000e+00 1.0835551e-12 8.4899246e-18
 7.4968040e-15 3.7670805e-10 2.6124387e-17 5.5974499e-11 6.8737841e-16]
 [1.8611118e-11 9.9971217e-01 7.6117715e-07 1.5593489e-13 2.3420353e-05
 2.1031763e-05 3.6960767e-08 9.0830480e-07 2.4174295e-04 9.0317899e-11]
 [1.0000000e+00 1.6215061e-17 2.6772470e-11 9.5272995e-16 1.2091799e-13
 9.2005847e-11 3.8800091e-10 1.9468100e-13 6.7435030e-10 1.0371038e-08]
 [4.1015207e-15 8.1935254e-22 2.6625205e-18 1.8816403e-17 1.0000000e+00
 6.2975496e-22 2.3692977e-17 1.9842730e-16 5.7110148e-16 1.3463005e-13]
 [3.7441495e-12 9.9992251e-01 7.4991817e-08 7.0841596e-12 2.2382910e-07
 3.2214842e-08 3.1644652e-11 6.7527244e-06 7.0347895e-05 3.5291223e-09]]
```

```
In [16]: print(numpy.argmax(prediction, axis=1))
print(Y_test[:6])
```

```
[7 2 1 0 4 1]
[[0. 0. 0. 0. 0. 0. 1. 0. 0.]
 [0. 0. 1. 0. 0. 0. 0. 0. 0.]
 [0. 1. 0. 0. 0. 0. 0. 0. 0.]
 [1. 0. 0. 0. 0. 0. 0. 0. 0.]
 [0. 0. 0. 0. 1. 0. 0. 0. 0.]
 [0. 1. 0. 0. 0. 0. 0. 0. 0.]]
```

Save the model

```
In [17]: model.save("Model.h5")
```

```
In [18]: model=load_model("model.h5")
```

```
In [19]: (X_train,y_train),(X_test,y_test)=mnist.load_data()
print('X_train:' +str(X_train.shape))
print('y_train:' +str(y_train.shape))
print('X_test:' +str(X_test.shape))
print('y_test:' +str(y_test.shape))
from matplotlib import pyplot
for i in range(9):
    pyplot.subplot(330+1+i)
    pyplot.imshow(X_train[i],cmap=pyplot.get_cmap('gray'))
    pyplot.show()
```

```
X_train:(60000, 28, 28)
y_train:(60000,)
X_test:(10000, 28, 28)
y_test:(10000,)
```

