22/08/2022, 10:49 Untitled7

## **NISHANT PAWAR IS 3**

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. The original creators of the database keep a list of some of the methods tested on it. In their original paper, they use a support-vector machine to get an error rate of 0.8%.

```
In [3]:
          import tensorflow
          import keras
          from tensorflow.keras import Sequential
          from keras.layers import Flatten, Dropout,Dense,Activation
          from tensorflow.keras.utils import to categorical
          from tensorflow.keras.datasets import mnist
In [4]:
          (X_train,y_train),(X_test,y_test) = mnist.load_data()
          X_train.shape,y_train.shape,X_test.shape,y_test.shape
         Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mn
         ist.npz
         11493376/11490434 [=====================] - 5s Ous/step
         11501568/11490434 [=============== ] - 5s @us/step
         ((60000, 28, 28), (60000,), (10000, 28, 28), (10000,))
Out[4]:
 In [5]:
          X_train = X_train.reshape(60000, 784)
          X_{\text{test}} = X_{\text{test.reshape}}(10000, 784)
          X_train = X_train.astype('float32')
          X_test = X_test.astype('float32')
In [6]:
          X_train /= 255
          X_test /= 255
In [7]:
          X_train.shape , X_test.shape
         ((60000, 784), (10000, 784))
Out[7]:
In [8]:
          from keras.utils import np utils
 In [9]:
          n classes = 10
          Y_train = np_utils.to_categorical(y_train, n_classes)
          Y test = np utils.to categorical(y test, n classes)
          Y train.shape
         (60000, 10)
Out[9]:
In [10]:
          X_train.shape, y_train.shape, X_test.shape, y_test.shape
         ((60000, 784), (60000,), (10000, 784), (10000,))
Out[10]:
```

```
model= Sequential()
In [11]:
In [12]:
          model.add(Dense(512, input shape=(784,)))
          model.add(Activation('relu'))
          model.add(Dropout(0.2))
          model.add(Dense(512))
          model.add(Activation('relu'))
          model.add(Dropout(0.2))
          model.add(Dense(10))
          model.add(Activation('softmax'))
In [13]:
          model.compile(loss='categorical_crossentropy', metrics=['accuracy'], optimizer='adam
In [14]:
          model.summary()
         Model: "sequential"
          Layer (type)
                                     Output Shape
                                                               Param #
          dense (Dense)
                                      (None, 512)
                                                               401920
          activation (Activation)
                                     (None, 512)
                                                               0
          dropout (Dropout)
                                     (None, 512)
                                     (None, 512)
          dense 1 (Dense)
                                                               262656
          activation_1 (Activation)
                                     (None, 512)
                                                               a
          dropout_1 (Dropout)
                                     (None, 512)
                                                               0
          dense 2 (Dense)
                                      (None, 10)
                                                               5130
          activation_2 (Activation)
                                     (None, 10)
         _____
         Total params: 669,706
         Trainable params: 669,706
         Non-trainable params: 0
In [15]:
          model.fit(X_train, Y_train,batch_size=128, epochs=20,verbose=2,validation_data=(X_te
         Epoch 1/20
         469/469 - 5s - loss: 0.2512 - accuracy: 0.9249 - val_loss: 0.0969 - val_accuracy: 0.
         9691 - 5s/epoch - 12ms/step
         Epoch 2/20
         469/469 - 4s - loss: 0.1000 - accuracy: 0.9689 - val_loss: 0.0924 - val_accuracy: 0.
         9707 - 4s/epoch - 9ms/step
         Epoch 3/20
         469/469 - 4s - loss: 0.0714 - accuracy: 0.9773 - val loss: 0.0673 - val accuracy: 0.
         9793 - 4s/epoch - 9ms/step
         Epoch 4/20
         469/469 - 4s - loss: 0.0554 - accuracy: 0.9823 - val_loss: 0.0656 - val_accuracy: 0.
         9791 - 4s/epoch - 8ms/step
         Epoch 5/20
```

469/469 - 3s - loss: 0.0457 - accuracy: 0.9857 - val\_loss: 0.0686 - val\_accuracy: 0.

22/08/2022, 10:49 Untitled7

```
9788 - 3s/epoch - 7ms/step
         Epoch 6/20
         469/469 - 4s - loss: 0.0399 - accuracy: 0.9868 - val_loss: 0.0628 - val_accuracy: 0.
         9814 - 4s/epoch - 8ms/step
         Epoch 7/20
         469/469 - 3s - loss: 0.0365 - accuracy: 0.9877 - val loss: 0.0705 - val accuracy: 0.
         9803 - 3s/epoch - 7ms/step
         Epoch 8/20
         469/469 - 4s - loss: 0.0319 - accuracy: 0.9890 - val_loss: 0.0726 - val_accuracy: 0.
         9802 - 4s/epoch - 8ms/step
         Epoch 9/20
         469/469 - 3s - loss: 0.0270 - accuracy: 0.9909 - val_loss: 0.0623 - val_accuracy: 0.
         9839 - 3s/epoch - 7ms/step
         Epoch 10/20
         469/469 - 4s - loss: 0.0261 - accuracy: 0.9911 - val_loss: 0.0649 - val_accuracy: 0.
         9827 - 4s/epoch - 7ms/step
         Epoch 11/20
         469/469 - 3s - loss: 0.0234 - accuracy: 0.9921 - val_loss: 0.0696 - val_accuracy: 0.
         9821 - 3s/epoch - 7ms/step
         Epoch 12/20
         469/469 - 3s - loss: 0.0200 - accuracy: 0.9933 - val_loss: 0.0694 - val_accuracy: 0.
         9811 - 3s/epoch - 7ms/step
         Epoch 13/20
         469/469 - 3s - loss: 0.0213 - accuracy: 0.9928 - val_loss: 0.0760 - val_accuracy: 0.
         9813 - 3s/epoch - 7ms/step
         Epoch 14/20
         469/469 - 3s - loss: 0.0209 - accuracy: 0.9933 - val_loss: 0.0699 - val_accuracy: 0.
         9818 - 3s/epoch - 7ms/step
         Epoch 15/20
         469/469 - 3s - loss: 0.0200 - accuracy: 0.9934 - val_loss: 0.0736 - val_accuracy: 0.
         9827 - 3s/epoch - 7ms/step
         Epoch 16/20
         469/469 - 3s - loss: 0.0172 - accuracy: 0.9945 - val_loss: 0.0687 - val_accuracy: 0.
         9844 - 3s/epoch - 7ms/step
         Epoch 17/20
         469/469 - 3s - loss: 0.0143 - accuracy: 0.9953 - val_loss: 0.0772 - val_accuracy: 0.
         9825 - 3s/epoch - 7ms/step
         Epoch 18/20
         469/469 - 3s - loss: 0.0180 - accuracy: 0.9938 - val loss: 0.0740 - val accuracy: 0.
         9845 - 3s/epoch - 7ms/step
         Epoch 19/20
         469/469 - 3s - loss: 0.0187 - accuracy: 0.9938 - val loss: 0.0637 - val accuracy: 0.
         9855 - 3s/epoch - 7ms/step
         Epoch 20/20
         469/469 - 4s - loss: 0.0141 - accuracy: 0.9953 - val_loss: 0.0708 - val_accuracy: 0.
         9836 - 4s/epoch - 8ms/step
         <keras.callbacks.History at 0x12090456c10>
Out[15]:
In [16]:
          model.save('model.h5')
```

## FINAL OUTPUT

22/08/2022, 10:49 Untitled7





