

```
In [ ]:

#!pip install np_utils

In [ ]:

#import tensorflow as tf
#print(tf.__version__)

In [ ]:

#!pip uninstall tensorflow
#!pip install tensorflow==2.2.0

In [ ]:

import pandas as pd
import tensorflow as tf
from tensorflow.keras.models import Sequential
from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Dense,Input,Conv2D,MaxPool2D,MaxPooling2D,Activation,Dropout,Flatten,BatchNormalization,MaxPool1D
from tensorflow.keras.optimizers import SGD,Adam
import numpy as np
```

Read the data

```
In [ ]:

train_data = pd.read_csv('train.csv')
train_data.head()
```

Out[]:

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	...	pixel774	pixel775	pixel776	pixel777	pixel778
0	1	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
2	1	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
3	4	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0

5 rows x 785 columns

```
In [ ]:

test_data = pd.read_csv('test.csv')
test_data.head()
```

Out[]:

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778
0	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0

5 rows x 784 columns

In []:

```
train_data['label'].value_counts()
```

Out[]:

```
1    4684
7    4401
3    4351
9    4188
2    4177
6    4137
0    4132
4    4072
8    4063
5    3795
Name: label, dtype: int64
```

Finding null values

In []:

```
def f1_null(train_data):
    count=train_data.isnull().sum().sum()
    return count
```

In []:

```
count_null = f1_null(train_data)
print(count_null)
```

0

To check the duplicates values

In []:

```
def f1_duplicates(train_data):
    count=train_data.duplicated().sum().sum()
    return count
```

In []:

```
count_duplicates = f1_duplicates(train_data)
print(count_duplicates)
```

0

In []:

```
print(train_data.shape)
print(test_data.shape)
```

```
(42000, 785)
(28000, 784)
```

Check the Column name for both train and test data set

In []:

```
def to_check_column(train_data):
    column_name = train_data.columns
    return column_name
```

In []:

```
train_column_name = to_check_column(train_data)
print("Column_name",train_column_name)
```

```
Column_name Index(['label', 'pixel0', 'pixel1', 'pixel2', 'pixel3', 'pixel4', 'pixel5',
                  'pixel6', 'pixel7', 'pixel8',
                  ...,
                  'pixel774', 'pixel775', 'pixel776', 'pixel777', 'pixel778', 'pixel779',
                  'pixel780', 'pixel781', 'pixel782', 'pixel783'],
                  dtype='object', length=785)
```

In []:

```
test_column_name = to_check_column(test_data)
print("Test column_name",test_column_name)
```

```
Test column_name Index(['pixel0', 'pixel1', 'pixel2', 'pixel3', 'pixel4', 'pixel5', 'pixel6',
                        'pixel7', 'pixel8', 'pixel9',
                        ...,
                        'pixel774', 'pixel775', 'pixel776', 'pixel777', 'pixel778', 'pixel779',
                        'pixel780', 'pixel781', 'pixel782', 'pixel783'],
                        dtype='object', length=784)
```

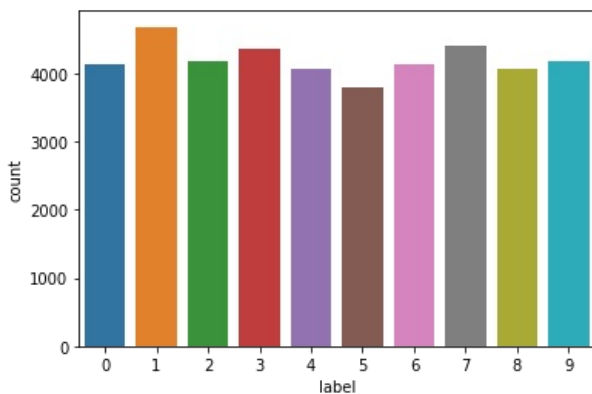
Count Plot

In []:

```
### let's use count plot to know the count of each label
def count_plot(train_data):
    sns.countplot(train_data.label)
```

In []:

```
count_plot(train_data)
```



Data is already clean. There is no duplicates and no missing values. Data is already balanced

Need to find range of pixels

In []:

```
train_data['pixel106'].value_counts()
```

Out[]:

```
0      41624
255      19
254      15
253       9
191       8
...
11        1
176        1
46         1
198        1
193        1
Name: pixel106, Length: 183, dtype: int64
```

In []:

```
train_data['pixel10'].value_counts()
```

Out[]:

```
0      42000
Name: pixel10, dtype: int64
```

As per observation, pixel intensities are currently between the range 0 and 255. we proceed to normalize the features between 0 and 1. Converted the numeric class vector to binary one hot encoding

```
In [ ]:

# Feature Normalization
y = train_data['label']
y.value_counts()

Out[ ]:

1    4684
7    4401
3    4351
9    4188
2    4177
6    4137
0    4132
4    4072
8    4063
5    3795
Name: label, dtype: int64
```

```
In [ ]:

print(type(y))

<class 'pandas.core.series.Series'>
```

```
In [ ]:

y_train = to_categorical(y, num_classes = 10)
```

```
In [ ]:

X = train_data.drop(labels=['label'],axis=1)
```

```
In [ ]:

X.head()
```

Out[]:

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778
0	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0

5 rows x 784 columns

Normalize the pixels values

```
In [ ]:

def normalize(data):
    data = data/255.0
    return data
```

```
In [ ]:

X = normalize(X)
print(X.shape)

(42000, 784)
```

In []:

```
X['pixel106'].value_counts()
```

Out[]:

```
0.000000    41624
1.000000      19
0.996078     15
0.992157      9
0.749020      8
...
0.043137      1
0.690196      1
0.180392      1
0.776471      1
0.756863      1
Name: pixel106, Length: 183, dtype: int64
```

Reshape the image

In []:

```
X_train = X
```

In []:

```
def image_reshape(X_train):
    X_train = X_train.values.reshape(-1,28,28,1)
    return X_train
```

In []:

```
X_train = image_reshape(X_train)
```

In []:

```
test_data = image_reshape(test_data)
```

In []:

```
print(X_train.shape)
print(test_data.shape)
```

```
(42000, 28, 28, 1)
(28000, 28, 28, 1)
```

Image has (28**28) pixels has been stock into Pandas. DataFrame as 1D Vectors of 784 values. We will reshape all data that indicates(height,width,channel).Channel 3 means RGB.

In []:

```
test_data = normalize(test_data)
print(test_data.shape)
```

```
(28000, 28, 28, 1)
```

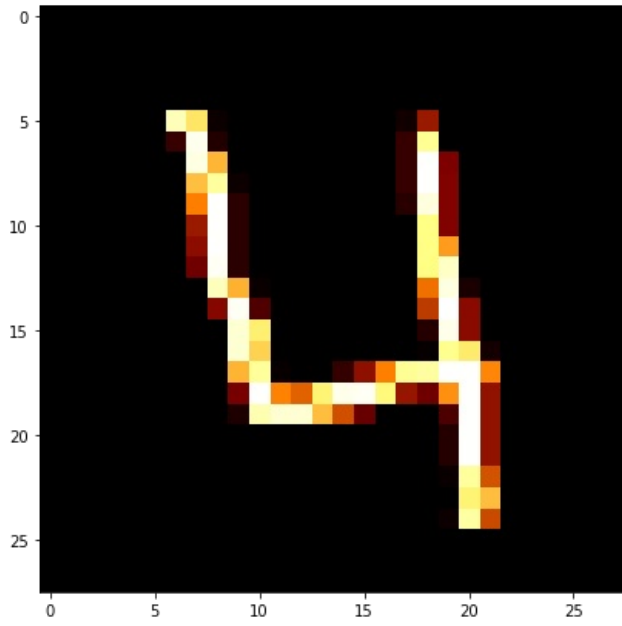
Let show the image

In []:

```
def image_show(X,idx):
    plt.figure(figsize=(7,7))
    grid_data = X.iloc[idx].to_numpy().reshape(28,28)
    plt.imshow(grid_data,interpolation=None,cmap='afmhot')
    plt.show()
```

In []:

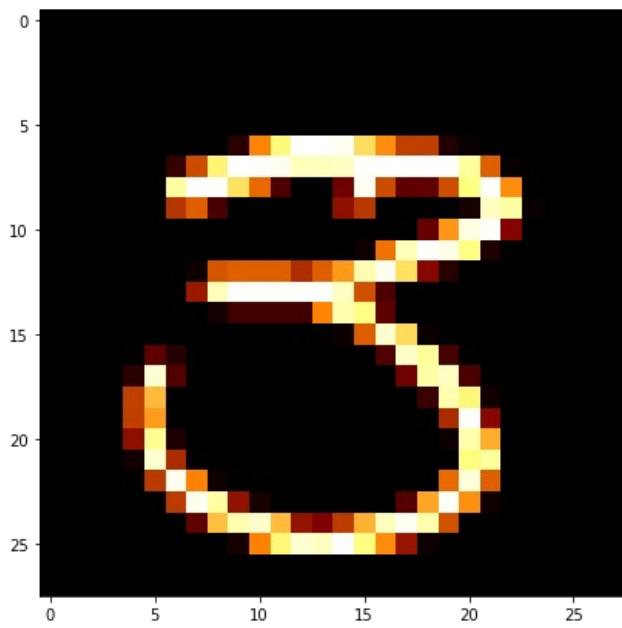
```
a1=image_show(X,3)
print(a1)
```



None

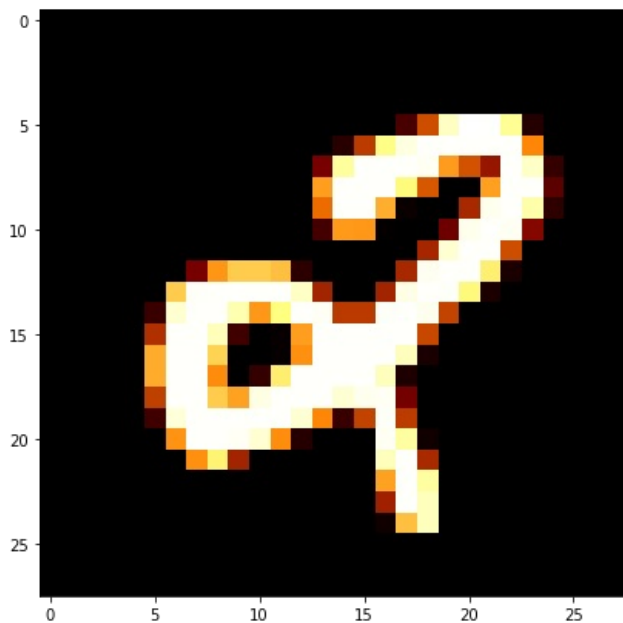
In []:

```
b1 = image_show(X,7)
b1
```



In []:

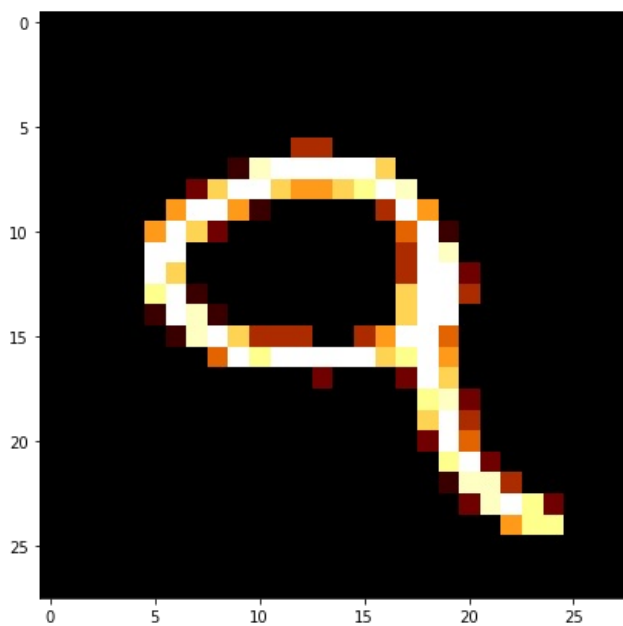
```
c1= image_show(X,104)
print(c1)
```



None

In []:

```
a2 = image_show(X,779)
print(a2)
```



None

Splitting train test image dataset

In []:

```
x_train,x_val,y_train,y_val = train_test_split(X_train,y_train,stratify=y_train,test_size=0.1,random_state=42)
print(x_train.shape)
print(x_val.shape)
print(y_train.shape)
print(y_val.shape)
```

(37800, 28, 28, 1)

(4200, 28, 28, 1)

(37800, 10)

(4200, 10)

Architecture of CNN -- $[(CNN2D-relu)_2 \rightarrow Maxpooling \rightarrow Dropout_2] \rightarrow Flatten \rightarrow Dense \rightarrow output$

In []:

```
model = Sequential()  
model.add(Conv2D(filters=32, kernel_size=(5,5), padding='same', activation='relu', input_shape=(28,28,1)))  
model.add(Conv2D(filters=32, kernel_size=(5,5), padding='same', activation='relu', input_shape=(28,28,1)))  
model.add(MaxPool2D(pool_size=(2,2)))  
model.add(Dropout(0.25))  
  
model.add(Conv2D(filters=64, kernel_size=(5,5), padding='same', activation='relu', input_shape=(28,28,1)))  
model.add(Conv2D(filters=64, kernel_size=(5,5), padding='same', activation='relu', input_shape=(28,28,1)))  
model.add(MaxPool2D(pool_size=(2,2)))  
model.add(Dropout(0.25))  
  
model.add(Flatten())  
model.add(Dense(512, activation='relu'))  
model.add(Dropout(0.25))  
model.add(Dense(10, activation='softmax'))
```

In []:

```
model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
=====		
conv2d_5 (Conv2D)	(None, 28, 28, 32)	832
conv2d_6 (Conv2D)	(None, 28, 28, 32)	25632
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 32)	0
dropout_5 (Dropout)	(None, 14, 14, 32)	0
conv2d_7 (Conv2D)	(None, 14, 14, 64)	51264
conv2d_8 (Conv2D)	(None, 14, 14, 64)	102464
max_pooling2d_4 (MaxPooling2D)	(None, 7, 7, 64)	0
dropout_6 (Dropout)	(None, 7, 7, 64)	0
flatten_2 (Flatten)	(None, 3136)	0
dense_5 (Dense)	(None, 512)	1606144
dropout_7 (Dropout)	(None, 512)	0
dense_6 (Dense)	(None, 10)	5130
=====		
Total params: 1,791,466		
Trainable params: 1,791,466		
Non-trainable params: 0		

In []:

```
# Define the optimizer  
optimizer = RMSprop(learning_rate=0.001, rho=0.9, momentum=0.1)
```

In []:

```
model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=['accuracy'])
```

In []:

```
print(y_train.shape)  
  
(37800, 10)
```


In []:

```
model.fit(x_train,y_train,epochs=2,batch_size=50,validation_data=(x_val,y_val))
```

Epoch 1/2

756/756 [=====] - 426s 563ms/step - loss: 0.1790 - accuracy: 0.9430 - val_loss: 0.0745 - val_accuracy: 0.9793

Epoch 2/2

756/756 [=====] - 426s 564ms/step - loss: 0.0554 - accuracy: 0.9840 - val_loss: 0.0443 - val_accuracy: 0.9881

Out[]:

<tensorflow.python.keras.callbacks.History at 0x7f91018a8d50>

Saved and load the Model

In []:

```
!mkdir -p saved_model
```

In []:

```
model.save(' ')
```

WARNING:tensorflow:From /usr/local/lib/python3.7/dist-packages/tensorflow/python/ops/resource_variable_ops.py:1817: calling BaseResourceVariable.__init__ (from tensorflow.python.ops.resource_variable_ops) with constraint is deprecated and will be removed in a future version.
Instructions for updating:
If using Keras pass *_constraint arguments to layers.

In []:

```
model.save('saved_model/my_model')
```

In []:

```
# my_model directory
!ls saved_model

# Contains an assets folder, saved_model.pb, and variables folder.
!ls saved_model/my_model

my_model
assets  saved_model.pb  variables
```

Reload a fresh keras model from saved model

In []:

```
#new_model = tf.keras.models.load_model('saved_model/my_model')

# Check its architecture
#new_model.summary()
```

Let's Do Confusion Matrix

In []:

```
#predict the values from validation data set
y_pred = model.predict(x_val)
#convert predictions classes to one hot vectors
y_pred_classes = np.argmax(y_pred,axis=1)
#convert validation observation to one hot vector
y_true = np.argmax(y_val,axis=1)
#calculate the confusion matrix
confusion_matrix(y_true,y_pred_classes)
```

Out[]:

```
array([[406, 0, 0, 0, 0, 0, 5, 0, 2, 0],
       [ 0, 463, 0, 0, 3, 0, 0, 0, 1, 1],
       [ 0, 2, 415, 0, 0, 0, 0, 0, 1, 0],
       [ 0, 0, 1, 428, 0, 1, 0, 3, 0, 2],
       [ 0, 0, 0, 0, 402, 0, 1, 0, 1, 3],
       [ 0, 0, 0, 2, 0, 374, 2, 0, 1, 1],
       [ 0, 0, 0, 0, 0, 0, 413, 0, 1, 0],
       [ 0, 0, 3, 0, 0, 0, 0, 435, 1, 1],
       [ 0, 0, 1, 0, 1, 0, 2, 0, 402, 0],
       [ 0, 0, 0, 0, 4, 0, 0, 0, 3, 412]])
```

In []:

```
classification_report(y_true,y_pred_classes)
```

Out[]:

```
'          precision    recall  f1-score   support\n\n 0          1.00      0.98      0.99         0
9          0.99      1.00      0.99        468\n
9          0.99      0.98      0.99        435\n
8          0.99      0.98      0.99        380\n
0.98      1.00      0.99      0.99        440\n
8          0.97      0.98      0.98        406\n
n accuracy      0.99      0.99      0.99        4200\n
4200\nweighted avg      0.99      0.99      0.99        4200\n'
```

In []:

```
y_predict = model.predict(test_data)

#select the index with maximum probability
results = np.argmax(y_predict,axis=1)
```

In []:

```
print(results)
```

[2 0 9 ... 3 9 2]

In []:

```
df_test = pd.read_csv("sample_submission.csv")
df_test.head()
```

Out[]:

	ImageId	Label
0	1	0
1	2	0
2	3	0
3	4	0
4	5	0

In []:

```
y_predict_lr_test = pd.DataFrame({"ImageId":df_test["ImageId"],"Label":results})
y_predict_lr_test.to_csv('mnist_digit_recognizer.csv', index=False)
y_predict_lr_test.head(5)
```

Out[]:

	ImageId	Label
0	1	2
1	2	0
2	3	9
3	4	9
4	5	3

[mnist_digit_recognizer.csv](#)

19 days ago by Anjali Sharma

[add submission details](#)

0.98475

Got the kaggle score 0.98

Apply image Augmented Technique

In []:

```
def image_augmentation(img,transform):
    'helper function to show the data augmentation'
    img = PIL.Image.open(img)
    fig,ax = plt.subplot(1,2,figsize=(2,4))
    ax[0].set_title(f'original image {img.size}')
    ax[0].image_show(img)
    ax[1].set_title(f'transformed image {img.size}')
    ax[1].tansform(img)
    ax[1].image_show(img)
```

In []:

```
#image_resize = transforms.Resize((23,45))
#image_augmentation(a1,image_resize)
```

In []:

```
print(x_train.shape)
print(y_val.shape)
```

```
(37800, 28, 28, 1)
(4200, 10)
```

Adding BatchNormalization which normalize the hidden layer

In []:

```
from tensorflow.python.keras.backend import batch_normalization
# Architecture of CNN -- [(CNN2D-relu)*2]->Maxpooling->Dropout]*2]->Flatten->Dense->output

model = Sequential()
model.add(Conv2D(filters=64,kernel_size=(5,5),padding='same',activation='relu',input_shape=(28,28,1)))
model.add(BatchNormalization(axis=1))
model.add(Conv2D(filters=64,kernel_size=(5,5),padding='same',activation='relu',input_shape=(28,28,1)))
model.add(BatchNormalization(axis=1))
model.add(MaxPool2D(pool_size=(2,2)))
model.add(BatchNormalization(axis=1))
model.add(Conv2D(filters=64,kernel_size=(5,5),padding='same',activation='relu',input_shape=(28,28,1)))
model.add(BatchNormalization(axis=1))
model.add(Conv2D(filters=64,kernel_size=(5,5),padding='same',activation='relu',input_shape=(28,28,1)))
model.add(BatchNormalization(axis=1))
model.add(MaxPool2D(pool_size=(2,2)))

model.add(Flatten())
model.add(BatchNormalization(axis=1))
model.add(Dense(512,activation='relu'))
model.add(BatchNormalization())
model.add(Dense(10,activation='softmax'))
```

In []:

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

In []:

```
model.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
=====		
conv2d_9 (Conv2D)	(None, 28, 28, 64)	1664
batch_normalization (BatchNo	(None, 28, 28, 64)	112
conv2d_10 (Conv2D)	(None, 28, 28, 64)	102464
batch_normalization_1 (Batch	(None, 28, 28, 64)	112
max_pooling2d_5 (MaxPooling2	(None, 14, 14, 64)	0
batch_normalization_2 (Batch	(None, 14, 14, 64)	56
conv2d_11 (Conv2D)	(None, 14, 14, 64)	102464
batch_normalization_3 (Batch	(None, 14, 14, 64)	56
conv2d_12 (Conv2D)	(None, 14, 14, 64)	102464
batch_normalization_4 (Batch	(None, 14, 14, 64)	56
max_pooling2d_6 (MaxPooling2	(None, 7, 7, 64)	0
flatten_3 (Flatten)	(None, 3136)	0
batch_normalization_5 (Batch	(None, 3136)	12544
dense_7 (Dense)	(None, 512)	1606144
batch_normalization_6 (Batch	(None, 512)	2048
dense_8 (Dense)	(None, 10)	5130
=====		
Total params: 1,935,314		
Trainable params: 1,927,822		
Non-trainable params: 7,492		

In []:

```
model.fit(x_train,y_train,epochs=10,batch_size=50,validation_data=(x_val,y_val))
```

```
Epoch 1/10
756/756 [=====] - 969s 1s/step - loss: 0.1237 - accuracy: 0.9620 - val_loss
: 0.0612 - val_accuracy: 0.9840
Epoch 2/10
756/756 [=====] - 964s 1s/step - loss: 0.0457 - accuracy: 0.9857 - val_loss
: 0.0439 - val_accuracy: 0.9890
Epoch 3/10
756/756 [=====] - 963s 1s/step - loss: 0.0341 - accuracy: 0.9891 - val_loss
: 0.0507 - val_accuracy: 0.9857
Epoch 4/10
756/756 [=====] - 960s 1s/step - loss: 0.0284 - accuracy: 0.9908 - val_loss
: 0.0367 - val_accuracy: 0.9900
Epoch 5/10
756/756 [=====] - 953s 1s/step - loss: 0.0237 - accuracy: 0.9927 - val_loss
: 0.0421 - val_accuracy: 0.9895
Epoch 6/10
756/756 [=====] - 959s 1s/step - loss: 0.0220 - accuracy: 0.9928 - val_loss
: 0.0343 - val_accuracy: 0.9924
Epoch 7/10
756/756 [=====] - 969s 1s/step - loss: 0.0209 - accuracy: 0.9930 - val_loss
: 0.0427 - val_accuracy: 0.9900
Epoch 8/10
756/756 [=====] - 963s 1s/step - loss: 0.0167 - accuracy: 0.9946 - val_loss
: 0.0540 - val_accuracy: 0.9876
Epoch 9/10
756/756 [=====] - 949s 1s/step - loss: 0.0174 - accuracy: 0.9947 - val_loss
: 0.0415 - val_accuracy: 0.9919
Epoch 10/10
391/756 [=====>.....] - ETA: 7:28 - loss: 0.0088 - accuracy: 0.9972
```

In []:

```
!mkdir -p saved_model_v2
```

In []:

```
model.save('saved_model_v2/my_model_v2')
```

In []:

```
#predict the values from validation set
y_pred = model.predict(x_val)
# convert predictions classes to one hot vectors
y_pred_classes = np.argmax(y_pred,axis=1)
#convert validation obserbation to one hot vecors
y_true = np.argmax(y_val,axis=1)
#let's call confusion matrix
confusion_matrix(y_true,y_pred_classes)
```

Out[]:

```
array([[412,    0,    0,    0,    0,    0,    1,    0,    0,    0],
       [  0, 467,    1,    0,    0,    0,    0,    0,    0,    0],
       [  0,    0, 416,    0,    0,    0,    0,    2,    0,    0],
       [  0,    0,    0, 433,    0,    1,    0,    0,    0,    1],
       [  0,    1,    0,    0, 402,    0,    0,    0,    0,    4],
       [  0,    0,    0,    5,    0, 366,    1,    0,    3,    5],
       [  0,    0,    0,    0,    0,    0, 412,    0,    2,    0],
       [  0,    0,    4,    1,    0,    0,    0, 435,    0,    0],
       [  3,    2,    0,    0,    0,    0,    2,    0, 399,    0],
       [  0,    1,    0,    0,    4,    0,    0,    1,    3, 410]])
```

In []:

```
y_predict = model.predict(test_data)
#let's take index with maximum probabilty
result = np.argmax(y_predict,axis=1)
print(result)
```

```
[2 0 9 ... 3 9 2]
```

In []:

```
df_test_batch = pd.read_csv("sample_submission.csv")
df_test_batch.head()
```

Out[]:

	ImageId	Label
0	1	0
1	2	0
2	3	0
3	4	0
4	5	0

In []:

```
y_predict_cnn_batch_test = pd.DataFrame({"ImageId": df_test_batch["ImageId"],"Label":result})
y_predict_cnn_batch_test.to_csv("mnist_data_set_submission_v2.csv",index=False)
y_predict_cnn_batch_test
```

Out[]:

	ImageId	Label
0	1	2
1	2	0
2	3	9
3	4	0
4	5	3
...
27995	27996	9
27996	27997	7
27997	27998	3
27998	27999	9
27999	28000	2

28000 rows × 2 columns

YOUR RECENT SUBMISSION



mnist_data_set_submission_v2.csv

Submitted by Anjali Sharma ·

Submitted 4 days ago

Score: 0.99103

Yaahh, Kaggle score is increased than previous one

saved the both Mode

In []:

```
!tar -czvf mnist.tar.gz ./saved_model
```

```
./saved_model/
./saved_model/my_model/
./saved_model/my_model/saved_model.pb
./saved_model/my_model/assets/
./saved_model/my_model/variables/
./saved_model/my_model/variables/variables.index
./saved_model/my_model/variables/variables.data-00000-of-00001
```

In []:

```
!tar -czvf mnist_v1.tar.gz ./saved_model_v2
```

```
./saved_model_v2/  
./saved_model_v2/my_model_v2/  
./saved_model_v2/my_model_v2/saved_model.pb  
./saved_model_v2/my_model_v2/assets/  
./saved_model_v2/my_model_v2/variables/  
./saved_model_v2/my_model_v2/variables/variables.index  
./saved_model_v2/my_model_v2/variables/variables.data-00000-of-00001
```

In []:

```
from keras.models import load_model
import tensorflow as tf
```

Load the MNIST Model -1

In []:

```
import tarfile
my_tar = tarfile.open('mnist.tar.gz')
my_tar.extractall()
my_tar.close()
```

Load MNIST MODEL -2

In []:

```
import tarfile
my_tar = tarfile.open('mnist_v1.tar.gz')
my_tar.extractall()
my_tar.close()
```

In []:

```
!ls saved_model/my_model
```

assets saved_model.pb variables

In []:

```
model_v1 = tf.keras.models.load_model('saved_model/my_model')
```

WARNING:tensorflow:SavedModel saved prior to TF 2.5 detected when loading Keras model. Please ensure that you are saving the model with model.save() or tf.keras.models.save_model(), *NOT* tf.saved_model.l.save(). To confirm, there should be a file named "keras_metadata.pb" in the SavedModel directory.

In []:

```
model_v1.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 28, 28, 32)	832
conv2d_1 (Conv2D)	(None, 28, 28, 32)	25632
max_pooling2d (MaxPooling2D)	(None, 14, 14, 32)	0
dropout (Dropout)	(None, 14, 14, 32)	0
conv2d_2 (Conv2D)	(None, 14, 14, 64)	51264
conv2d_3 (Conv2D)	(None, 14, 14, 64)	102464
max_pooling2d_1 (MaxPooling2D)	(None, 7, 7, 64)	0
dropout_1 (Dropout)	(None, 7, 7, 64)	0
flatten (Flatten)	(None, 3136)	0
dense (Dense)	(None, 512)	1606144
dropout_2 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 10)	5130
=====		
Total params: 1,791,466		
Trainable params: 1,791,466		
Non-trainable params: 0		

In []:

```
!ls saved_model_v2/
```

my_model_v2

In []:

```
!ls saved_model_v2/my_model_v2/
```

assets saved_model.pb variables

In []:

```
import warnings
warnings.filterwarnings("ignore")
model_v2 = tf.keras.models.load_model('saved_model_v2/my_model_v2')
```

WARNING:tensorflow:SavedModel saved prior to TF 2.5 detected when loading Keras model. Please ensure that you are saving the model with model.save() or tf.keras.models.save_model(), *NOT* tf.saved_model.save(). To confirm, there should be a file named "keras_metadata.pb" in the SavedModel directory.

In []:

```
model_v2.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
=====		
conv2d_9 (Conv2D)	(None, 28, 28, 64)	1664
batch_normalization (Batch Normalization)	(None, 28, 28, 64)	112
conv2d_10 (Conv2D)	(None, 28, 28, 64)	102464
batch_normalization_1 (Batch Normalization)	(None, 28, 28, 64)	112
max_pooling2d_5 (MaxPooling2D)	(None, 14, 14, 64)	0
batch_normalization_2 (Batch Normalization)	(None, 14, 14, 64)	56
conv2d_11 (Conv2D)	(None, 14, 14, 64)	102464
batch_normalization_3 (Batch Normalization)	(None, 14, 14, 64)	56
conv2d_12 (Conv2D)	(None, 14, 14, 64)	102464
batch_normalization_4 (Batch Normalization)	(None, 14, 14, 64)	56
max_pooling2d_6 (MaxPooling2D)	(None, 7, 7, 64)	0
flatten_3 (Flatten)	(None, 3136)	0
batch_normalization_5 (Batch Normalization)	(None, 3136)	12544
dense_7 (Dense)	(None, 512)	1606144
batch_normalization_6 (Batch Normalization)	(None, 512)	2048
dense_8 (Dense)	(None, 10)	5130
=====		
Total params: 1,935,314		
Trainable params: 1,927,822		
Non-trainable params: 7,492		

In []:

```
!pip install -q streamlit
```

In []:

```
!pip install -q pyngrok
```

In []:

```
!pip install -q streamlit_ace
```

REFERENCE: <https://www.analyticsvidhya.com/blog/2020/12/deploying-machine-learning-models-using-streamlit-an-introductory-guide-to-model-deployment/> (<https://www.analyticsvidhya.com/blog/2020/12/deploying-machine-learning-models-using-streamlit-an-introductory-guide-to-model-deployment/>)

In []:

```
import pandas as pd
import streamlit as st
```

In []:

```
test_data = pd.read_csv("test.csv")
```

In []:

```
test_data.head()
```

Out[]:

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778
0	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0

5 rows x 784 columns

In []:

```
test_data.shape
```

Out[]:

```
(28000, 784)
```

In []:

```
test_data['pixel1'].value_counts()
```

Out[]:

```
0    28000
Name: pixel1, dtype: int64
```

In []:

```
import matplotlib.pyplot as plt
import numpy as np
```

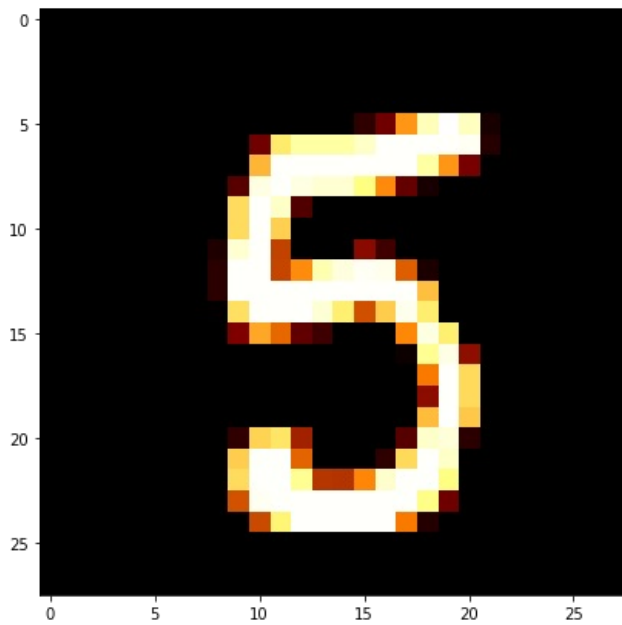
In []:

```
def image_show(X,idx):
    plt.figure(figsize=(7,7))
    grid_data = X.iloc[idx].to_numpy().reshape(28,28)
    plt.imshow(grid_data,interpolation=None,cmap='afmhot')
    plt.show()
```

In []:

```
test_data = test_data/255.0
```

```
image_show(test_data,10)
```



In []:

```
def prediction(idx):  
    plt.figure(figsize=(7,7))  
    image= test_data.iloc[idx].to_numpy().reshape(1,28,28)  
    p = model_v1.predict(image)  
    return p
```

In []:

```
p = prediction(9)  
print("prediction {}".format(np.argmax(p)))
```

prediction 3

<Figure size 504x504 with 0 Axes>

Model Deployment of MNIST DATA set using streamlit library

In []:

```
%%writefile app.py
import streamlit as st
import matplotlib.pyplot as plt
from keras.models import load_model
import tensorflow as tf
import numpy as np
import pandas as pd
model_v1 = tf.keras.models.load_model('saved_model/my_model')
test_data = pd.read_csv("test.csv")
def prediction(idx):
    idx = int(idx)
    plt.figure(figsize=(7,7))
    image= test_data.iloc[idx].to_numpy().reshape(1,28,28)
    p = model_v1.predict(image)
    return p
# this is main function which defines the web pages
def main():
    #front end element on web pages
    html_temp = """
    <div style = "background-color:yellow;padding:13px">
    <h1 style = "color:black;text-align:center;">Digit Recogintion</h1>
    </div>
    """
    # display the front end aspect
    st.markdown(html_temp, unsafe_allow_html = True)
    idx = st.number_input("Enter the MNIST image index")
    result= " "
    if st.button("Predict"):
        result = prediction(idx)
        st.success("MNIST digit Recogintion based on entered image's index = {}".format(np.argmax(result)))
if __name__ == '__main__':
    main()
```

Overwriting app.py

In []:

```
!streamlit run app.py &>/dev/null&
```

In []:

```
!ngrok authtoken 24QK9tsxhgVhAIssPQhPrFNQSWg_3j6PtXWiKHCUnCtXn6BbG
```

Authtoken saved to configuration file: /root/.ngrok2/ngrok.yml

In []:

```
from pyngrok import ngrok
public_url = ngrok.connect('8501')
public_url
```

```
INFO:pyngrok.ngrok:Opening tunnel named: http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7
2022-09-12 19:03:11.716 INFO    pyngrok.ngrok: Opening tunnel named: http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:11+0000 lvl=info msg="no configuration paths supplied"
2022-09-12 19:03:11.860 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:11+0000 lvl=info msg="no configuration paths supplied"
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:11+0000 lvl=info msg="using configuration at default config path" path=/root/.ngrok2/ngrok.yml
2022-09-12 19:03:11.868 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:11+0000 lvl=info msg="using configuration at default config path" path=/root/.ngrok2/ngrok.yml
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:11+0000 lvl=info msg="open config file" path=/root/.ngrok2/ngrok.yml err=nil
2022-09-12 19:03:11.876 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:11+0000 lvl=info msg="open config file" path=/root/.ngrok2/ngrok.yml err=nil
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:11+0000 lvl=info msg="starting web service" obj=web addr=127.0.0.1:4040
2022-09-12 19:03:11.884 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:11+0000 lvl=info msg="starting web service" obj=web addr=127.0.0.1:4040
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg="tunnel session started" obj=tunnels.session
2022-09-12 19:03:12.031 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg="tunnel session started" obj=tunnels.session
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg="client session established" obj=csess id=ba6443affdb0
2022-09-12 19:03:12.041 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg="client session established" obj=csess id=ba6443affdb0
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=25dalacf266fdeae
2022-09-12 19:03:12.069 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=25dalacf266fdeae
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=25dalacf266fdeae status=200 dur=460.431µs
2022-09-12 19:03:12.082 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=25dalacf266fdeae status=200 dur=460.431µs
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=8c31481f2f4089b1
2022-09-12 19:03:12.097 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=8c31481f2f4089b1
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=8c31481f2f4089b1 status=200 dur=132.412µs
2022-09-12 19:03:12.103 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=8c31481f2f4089b1 status=200 dur=132.412µs
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=c49770216db854d3
2022-09-12 19:03:12.109 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=start pg=/api/tunnels id=c49770216db854d3
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg="started tunnel" obj=tunnels name="http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7 (http)" addr=http://localhost:8501 url=http://58b0-34-138-217-149.ngrok.io
2022-09-12 19:03:12.160 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg="started tunnel" obj=tunnels name="http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7 (http)" addr=http://localhost:8501 url=http://58b0-34-138-217-149.ngrok.io
```

Out []:

```
<NgrokTunnel: "http://58b0-34-138-217-149.ngrok.io" -> "http://localhost:8501">
```

```
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg="started tunnel" obj=tunnels name="http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7" addr=http://localhost:8501 url=https://58b0-34-138-217-149.ngrok.io
2022-09-12 19:03:12.167 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg="started tunnel" obj=tunnels name="http-8501-6a1477c9-ea55-4714-b1d7-68b961dd0ea7" addr=http://localhost:8501 url=https://58b0-34-138-217-149.ngrok.io
INFO:pyngrok.process.ngrok:t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=c49770216db854d3 status=201 dur=90.402484ms
2022-09-12 19:03:12.176 INFO    pyngrok.process.ngrok: t=2022-09-12T19:03:12+0000 lvl=info msg=end pg=/api/tunnels id=c49770216db854d3 status=201 dur=90.402484ms
```