

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

In [2]: train_data = pd.read_csv('train.csv')
test_data = pd.read_csv('test.csv')

In [3]: train_data.head()

Out[3]:
```

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
0	1	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	0
2	1	0	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	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0

5 rows x 785 columns

```
In [4]: train_data.tail()

Out[4]:
```

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
41995	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
41996	1	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
41997	7	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
41998	6	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
41999	9	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0

5 rows x 785 columns

```
In [5]: test_data.tail()

Out[5]:
```

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
27995	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
27996	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
27997	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
27998	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0
27999	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0

5 rows x 784 columns

```
In [6]: test_data.head()

Out[6]:
```

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
0	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	0
2	0	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	0
4	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0

5 rows x 784 columns

```
In [7]: train_data.shape

Out[7]: (42000, 785)

In [8]: test_data.shape

Out[8]: (28000, 784)

In [9]: train_data.columns

Out[9]: Index(['label', 'pixel0', 'pixel1', 'pixel2', 'pixel3', 'pixel4', 'pixel5',
        'pixel6', 'pixel7', 'pixel8', 'pixel9', ..., 'pixel774', 'pixel775', 'pixel776', 'pixel777', 'pixel778', 'pixel779',
        'pixel780', 'pixel781', 'pixel782', 'pixel783'],
        dtype='object', length=785)

In [10]: test_data.columns

Out[10]: 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)

In [11]: train_data.duplicated().sum()

Out[11]: 0

In [12]: test_data.duplicated().sum()

Out[12]: 0

In [13]: train_data.isnull().sum()

Out[13]: label      0
pixel0      0
pixel1      0
pixel2      0
pixel3      0
pixel4      0
pixel5      0
pixel6      0
pixel7      0
pixel8      0
pixel9      0
...
pixel774    0
pixel775    0
pixel776    0
pixel777    0
pixel778    0
pixel779    0
pixel780    0
pixel781    0
pixel782    0
pixel783    0
Length: 785, dtype: int64

In [14]: test_data.isnull().sum()

Out[14]: pixel0      0
pixel1      0
pixel2      0
pixel3      0
pixel4      0
pixel5      0
pixel6      0
pixel7      0
pixel8      0
pixel9      0
...
pixel774    0
pixel775    0
pixel776    0
pixel777    0
pixel778    0
pixel779    0
pixel780    0
pixel781    0
pixel782    0
pixel783    0
Length: 784, dtype: int64

In [15]: train_data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 42000 entries, 0 to 41999
Columns: 785 entries, label to pixel783
dtypes: int64(785)
memory usage: 251.5 MB

In [16]: test_data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 28000 entries, 0 to 27999
Columns: 784 entries, pixel0 to pixel783
dtypes: int64(784)
memory usage: 167.5 MB

In [17]: train_data.describe()

Out[17]:
```

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
count	42000.000000	42000.0	42000.0	42000.0	42000.0	42000.0	42000.0	42000.0	42000.0	42000.0	...	42000.000000	42000.000000	42000.000000	42000.000000	42000.000000	42000.000000
mean	4.456643	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	0.111775
std	2.887730	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	4.632144
min	0.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	2.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
50%	4.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
75%	7.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
max	9.000000	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	254.000000	254.000000	254.000000	254.000000	254.000000	254.000000

8 rows x 785 columns

```
In [18]: test_data.describe()

Out[18]:
```

	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel7	pixel8	pixel9	...	pixel774	pixel775	pixel776	pixel777	pixel778	pixel779
count	28000.0	28000.0	28000.0	28000.0	28000.0	28000.0	28000.0	28000.0	28000.0	28000.0	...	28000.000000	28000.000000	28000.000000	28000.000000	28000.000000	28000.000000
mean	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.164607	0.073214	0.164607	0.073214	0.164607	0.073214
std	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	5.473293	3.616811	5.473293	3.616811	5.473293	3.616811
min	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
50%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
75%	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
max	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	253.000000	254.000000	253.000000	254.000000	253.000000	254.000000

8 rows x 784 columns

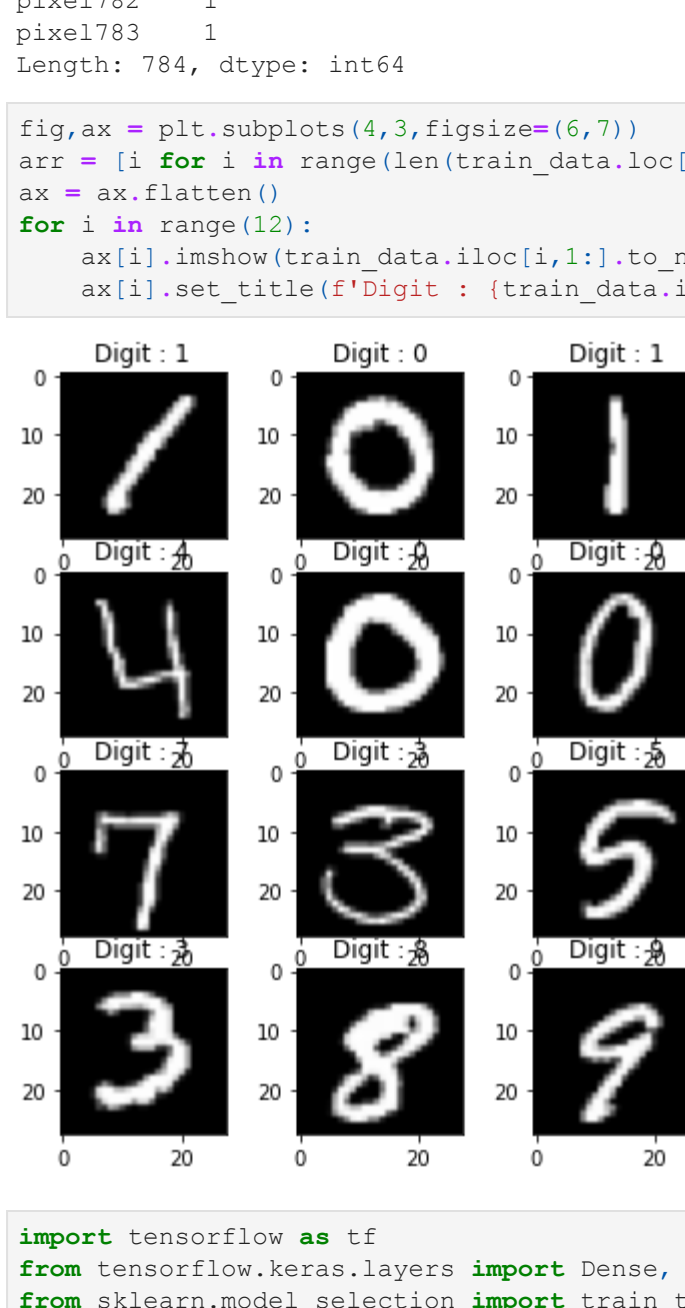
```
In [19]: train_data.nunique()

Out[19]: label      10
pixel0      1
pixel1      1
pixel2      1
pixel3      1
pixel4      1
pixel5      1
pixel6      1
pixel7      1
pixel8      1
pixel9      1
...
pixel774    1
pixel775    1
pixel776    1
pixel777    1
pixel778    1
pixel779    1
pixel780    1
pixel781    1
pixel782    1
pixel783    1
Length: 785, dtype: int64

In [20]: test_data.nunique()

Out[20]: pixel0      1
pixel1      1
pixel2      1
pixel3      1
pixel4      1
pixel5      1
pixel6      1
pixel7      1
pixel8      1
pixel9      1
...
pixel774    1
pixel775    1
pixel776    1
pixel777    1
pixel778    1
pixel779    1
pixel780    1
pixel781    1
pixel782    1
pixel783    1
Length: 784, dtype: int64


In [22]: fig,ax = plt.subplots(4,3,figsize=(6,7))
arr = [i for i in range(len(train_data.loc[:, 'label'].unique()))]
ax = ax.flatten()
for i in range(12):
    ax[i].imshow(train_data.iloc[i,1:].to_numpy().reshape((28,28)), cmap='gray')
    ax[i].set_title(f'Digit : {train_data.iloc[i,0]}')
```



```
In [23]: import tensorflow as tf
import tensorflow.keras.layers import Dense, Conv2D, MaxPooling2D, Dropout, Flatten
from sklearn.model_selection import train_test_split

In [24]: input_data= train_data.drop(['label'],axis=1)
target= train_data.label

In [25]: plt.figure(figsize=(15,8))
sns.countplot(x=train_data.loc[:, 'label'], data=train_data.loc[:, 'label'].dropna())
plt.xticks(rotation=90)
plt.show()
```



```
In [26]: x_train,x_test,y_train,y_test = train_test_split(input_data,target,test_size=0.3)

In [27]: y_train_enc = tf.keras.utils.to_categorical(y_train)
y_test_enc = tf.keras.utils.to_categorical(y_test)

In [28]: y_train_enc[0]

Out[28]: array([0., 0., 0., 0., 0., 0., 0., 1., 0., 0.], dtype=float32)

In [29]: x_train = x_train.values.reshape((-1,28,28,1))
x_test = x_test.values.reshape((-1,28,28,1))

In [31]: def model():
nn = tf.keras.models.Sequential()
nn.add(Conv2D(64,kernel_size=(3,3),input_shape=(28,28,1),activation='relu'))
nn.add(Conv2D(64,kernel_size=(3,3),activation='relu'))
nn.add(MaxPooling2D(pool_size=(2,2)))
nn.add(Dropout(0.2))
nn.add(Conv2D(128,kernel_size=(3,3),activation='relu'))
nn.add(Conv2D(128,kernel_size=(3,3),activation='relu'))
nn.add(MaxPooling2D(pool_size=(2,2)))
nn.add(Dropout(0.2))
nn.add(Conv2D(256,kernel_size=(3,3),activation='relu'))
nn.add(Flatten())
nn.add(Dense(10,activation='softmax'))
nn.compile(optimizer='Adam',loss='categorical_crossentropy',metrics=['accuracy'])
return nn

In [32]: model = model()
model.summary()

Model: "sequential"

Layer (type) Output Shape Param #
-----
conv2d (Conv2D) (None, 26, 26, 64) 640
conv2d_1 (Conv2D) (None, 24, 24, 64) 36928
max_pooling2d (MaxPooling2D) (None, 12, 12, 64) 0
dropout (Dropout) (None, 12, 12, 64) 0
conv2d_2 (Conv2D) (None, 10, 10, 128) 73856
conv2d_3 (Conv2D) (None, 8, 8, 128) 147584
max_pooling2d_1 (MaxPooling2D) (None, 4, 4, 128) 0
dropout_1 (Dropout) (None, 4, 4, 128) 0
conv2d_4 (Conv2D) (None, 2, 2, 256) 295168
flatten (Flatten) (None, 1024) 0
dense (Dense) (None, 10) 10250
Total params: 564,426
Trainable params: 564,426
Non-trainable params: 0

2022-08-03 19:24:47.839214: I tensorflow/core/platform/cpu_feature_guard.cc:151] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 AVX512F FMA
To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

In [33]: model.fit(x_train,y_train_enc,epochs=20)

Epoch 1/20
919/919 [=====] - 66s 71ms/step - loss: 0.3240 - accuracy: 0.9159
Epoch 2/20
919/919 [=====] - 67s 73ms/step - loss: 0.0792 - accuracy: 0.9756
Epoch 3/20
919/919 [=====] - 65s 70ms/step - loss: 0.0613 - accuracy: 0.9810
Epoch 4/20
919/919 [=====] - 67s 73ms/step - loss: 0.0517 - accuracy: 0.9834
Epoch 5/20
919/919 [=====] - 67s 73ms/step - loss: 0.0422 - accuracy: 0.9875
Epoch 6/20
919/919 [=====] - 64s 69ms/step - loss: 0.0451 - accuracy: 0.9854
Epoch 7/20
919/919 [=====] - 66s 72ms/step - loss: 0.0380 - accuracy: 0.9883
Epoch 8/20
919/919 [=====] - 66s 72ms/step - loss: 0.0325 - accuracy: 0.9902
Epoch 9/20
919/919 [=====] - 66s 72ms/step - loss: 0.0325 - accuracy: 0.9902
Epoch 10/20
919/919 [=====] - 71s 77ms/step - loss: 0.0319 - accuracy: 0.9909
Epoch 11/20
919/919 [=====] - 75s 81ms/step - loss: 0.0303 - accuracy: 0.9907
Epoch 12/20
919/919 [=====] - 70s 77ms/step - loss: 0.0242 - accuracy: 0.9927
Epoch 13/20
919/919 [=====] - 70s 77ms/step - loss: 0.0282 - accuracy: 0.9924
Epoch 14/20
919/919 [=====] - 71s 77ms/step - loss: 0.0295 - accuracy: 0.9915
Epoch 15/20
919/919 [=====] - 71s 77ms/step - loss: 0.0275 - accuracy: 0.9927
Epoch 16/20
919/919 [=====] - 70s 76ms/step - loss: 0.0239 - accuracy: 0.9927
Epoch 17/20
919/919 [=====] - 70s 77ms/step - loss: 0.0264 - accuracy: 0.9932
Epoch 18/20
919/919 [=====] - 69s 76ms/step - loss: 0.0270 - accuracy: 0.9926
Epoch 19/20
919/919 [=====] - 70s 77ms/step - loss: 0.0210 - accuracy: 0.9939
Epoch 20/20
919/919 [=====] - 77s 83ms/step - loss: 0.0260 - accuracy: 0.9930
Out[33]: <keras.callbacks.History at 0x7fc523c1c430>

In [34]: loss, acc = model.evaluate(x_test,y_test_enc)

394/394 [=====] - 6s 14ms/step - loss: 0.0732 - accuracy: 0.9891

In [35]: model.summary()

Model: "sequential"

Layer (type) Output Shape Param #
-----
conv2d (Conv2D) (None, 26, 26, 64) 640
conv2d_1 (Conv2D) (None, 24, 24, 64) 36928
max_pooling2d (MaxPooling2D) (None, 12, 12, 64) 0
dropout (Dropout) (None, 12, 12, 64) 0
conv2d_2 (Conv2D) (None, 10, 10, 128) 73856
conv2d_3 (Conv2D) (None, 8, 8, 128) 147584
max_pooling2d_1 (MaxPooling2D) (None, 4, 4, 128) 0
dropout_1 (Dropout) (None, 4, 4, 128) 0
conv2d_4 (Conv2D) (None, 2, 2, 256) 295168
flatten (Flatten) (None, 1024) 0
dense (Dense) (None, 10) 10250
Total params: 564,426
Trainable params: 564,426
Non-trainable params: 0

In [36]: print('Loss = ',loss)
print('Accuracy = ',acc*100)

Loss = 0.07320599257946014
Accuracy = 98.9126980304718

In [37]: pred = model.predict(test_data.values.reshape((-1,28,28,1)))
pred.shape

Out[37]: (28000, 10)

In [39]: preds = []
for i in range(28000):
    preds.append(np.argmax(pred[i]))

In [40]: sub = pd.read_csv('sample_submission.csv')
sub['Label'] = preds

In [41]: sub

Out[41]:
```

	ImageId	Label
0	1	0
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 x 2 columns

```
In [42]: sub.to_csv('submission.csv',index=False)

In [ ]:
```