

```
#Importing packages
import tensorflow as tf
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
```

▼ Loading dataset

```
#loading dataset
(x_train,y_train),(x_test,y_test)=tf.keras.datasets.mnist.load_data()
```

```
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mni
11493376/11490434 [=====] - 11s 1us/step
11501568/11490434 [=====] - 11s 1us/step
```



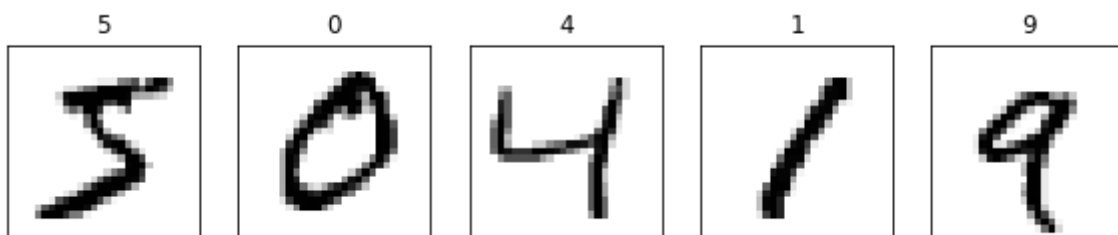
▼ Plotting count plot

```
plt.figure(figsize = (10,8))
sns.countplot(y_train)
```

C:\Users\Nishant\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning
warnings.warn(

▼ Displaying some images

```
#Dataset properties
# Display some images
fig, axes = plt.subplots(ncols=5, sharex=False,
                          sharey=True, figsize=(10, 4))
for i in range(5):
    axes[i].set_title(y_train[i])
    axes[i].imshow(x_train[i], cmap='gray_r')
    axes[i].get_xaxis().set_visible(False)
    axes[i].get_yaxis().set_visible(False)
plt.show()
```



▼ Pre-processing the data

```
# Pre-processing the data
print('Training images shape : ',x_train.shape)
print('Testing images shape : ',x_test.shape)
```

```
Training images shape : (60000, 28, 28)
Testing images shape : (10000, 28, 28)
```

```
x_train = x_train.reshape(x_train.shape[0], 28, 28, 1)
x_test = x_test.reshape(x_test.shape[0], 28, 28, 1)
input_shape = (28, 28, 1)
```

```
#applying normalization
x_train=x_train/255.0
x_testg=x_test/255.0
num_classes = 10
```

▼ Creating the model

Model Architecture

We will have to first build the model architecture and define it based on our dataset. We are going to add the following layers:

1. Conv2D - for the convolution layers
2. Dropout - to prevent overfitting
3. Dense - a fully connected layer
4. Softmax activation - This is used to convert all predictions into probability The model architecture can be tuned to get optimal performance

so i am going to create a model with

CNN + Three layers + relu + (3,3) kernel_size + Dropout rate (0.3)

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, Flatten, Activation
from tensorflow.keras.layers import Conv2D, MaxPooling2D
from tensorflow.keras.layers import BatchNormalization
```

```
model = Sequential()
```

```
model.add(Conv2D(128, kernel_size=(3, 3),
                activation=tf.nn.relu,
                input_shape=input_shape))
model.add(BatchNormalization())
model.add(Dropout(0.3))
```

```
model.add(Conv2D(64, (3, 3), activation=tf.nn.relu))
model.add(BatchNormalization())
model.add(Dropout(0.3))
```

```
model.add(Conv2D(32, (3, 3), activation=tf.nn.relu))
model.add(BatchNormalization())
model.add(Dropout(0.3))
```

```
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.3))
model.add(Flatten())
model.add(Dense(128, activation=tf.nn.relu))
model.add(Dropout(0.3))
model.add(Dense(num_classes, activation=tf.nn.softmax))
```

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 26, 26, 128)	1280

batch_normalization (Batch Normalization)	(None, 26, 26, 128)	512
dropout (Dropout)	(None, 26, 26, 128)	0
conv2d_1 (Conv2D)	(None, 24, 24, 64)	73792
batch_normalization_1 (Batch Normalization)	(None, 24, 24, 64)	256
dropout_1 (Dropout)	(None, 24, 24, 64)	0
conv2d_2 (Conv2D)	(None, 22, 22, 32)	18464
batch_normalization_2 (Batch Normalization)	(None, 22, 22, 32)	128
dropout_2 (Dropout)	(None, 22, 22, 32)	0
max_pooling2d (MaxPooling2D)	(None, 11, 11, 32)	0
dropout_3 (Dropout)	(None, 11, 11, 32)	0
flatten (Flatten)	(None, 3872)	0
dense (Dense)	(None, 128)	495744
dropout_4 (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 10)	1290

=====

Total params: 591,466
 Trainable params: 591,018
 Non-trainable params: 448

▼ Training the model

```
# Train the model
model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
history=model.fit(x=x_train,
                  y=y_train,
                  validation_split=0.1,
                  epochs=10)
```

```
Epoch 1/10
1688/1688 [=====] - 572s 321ms/step - loss: 0.2633 - accuracy: 0.0000
Epoch 2/10
1688/1688 [=====] - 551s 326ms/step - loss: 0.1112 - accuracy: 0.0000
Epoch 3/10
1688/1688 [=====] - 553s 328ms/step - loss: 0.0897 - accuracy: 0.0000
```

```

Epoch 4/10
1688/1688 [=====] - 547s 324ms/step - loss: 0.0750 - accuracy: 0.9418
Epoch 5/10
1688/1688 [=====] - 546s 323ms/step - loss: 0.0644 - accuracy: 0.9514
Epoch 6/10
1688/1688 [=====] - 544s 322ms/step - loss: 0.0597 - accuracy: 0.9571
Epoch 7/10
1688/1688 [=====] - 587s 348ms/step - loss: 0.0534 - accuracy: 0.9628
Epoch 8/10
1688/1688 [=====] - 632s 374ms/step - loss: 0.0480 - accuracy: 0.9685
Epoch 9/10
1688/1688 [=====] - 632s 374ms/step - loss: 0.0441 - accuracy: 0.9742
Epoch 10/10
1688/1688 [=====] - 632s 374ms/step - loss: 0.0418 - accuracy: 0.9799

```

▼ Saving and loading the model

```
model.save('MNproject.h5')
```

```
from tensorflow.keras.models import load_model
model = load_model('MNproject.h5')
```

▼ Evaluating the model

```
# Evaluate the model
loss_and_acc=model.evaluate(x_test,y_test)
print("Test Loss", loss_and_acc[0])
print("Test Accuracy", loss_and_acc[1])
```

```

313/313 [=====] - 20s 62ms/step - loss: 2.5056 - accuracy: 0.9876
Test Loss 2.505582523345947
Test Accuracy 0.9876999855041504

```

▼ Plotting Training & Validation plots

```

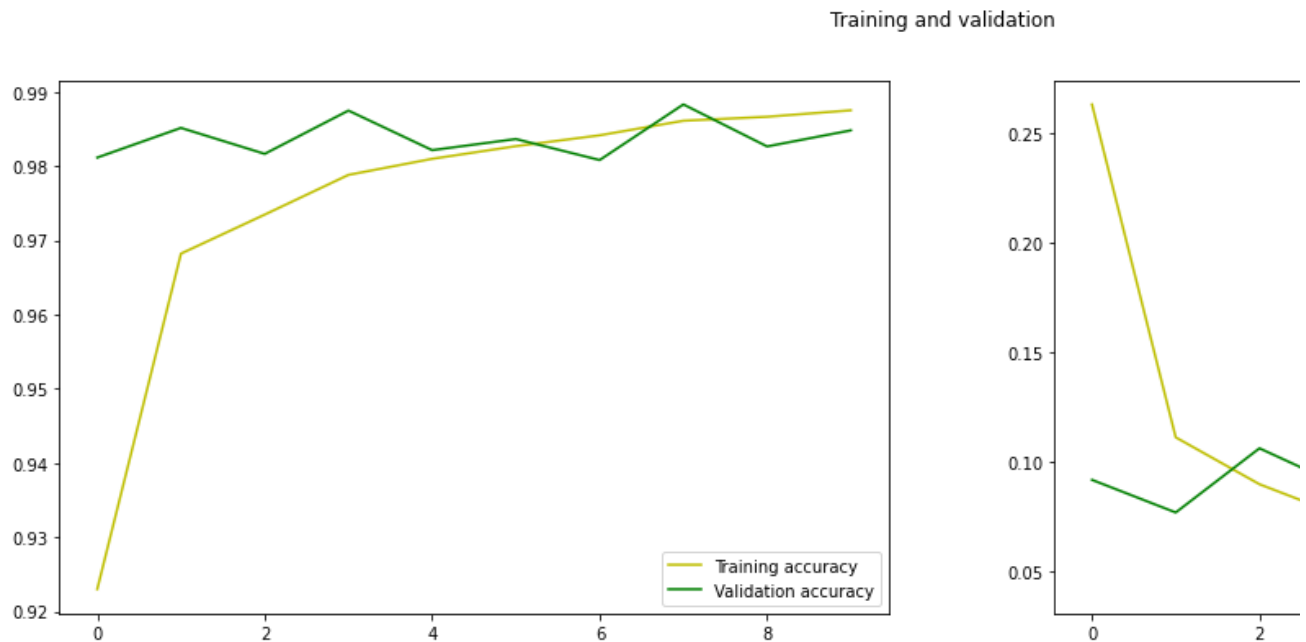
acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(len(acc))

fig, ax = plt.subplots(nrows=1, ncols=2, figsize=(20, 6))
ax[0].plot(epochs, acc, 'y', label='Training accuracy')
ax[0].plot(epochs, val_acc, 'g', label='Validation accuracy')
ax[0].legend(loc=0)
ax[1].plot(epochs, loss, 'y', label='Training loss')

```

```
ax[1].plot(epochs, val_loss, 'g', label='Validation loss')
ax[1].legend(loc=0)
```

```
plt.suptitle('Training and validation')
plt.show()
```



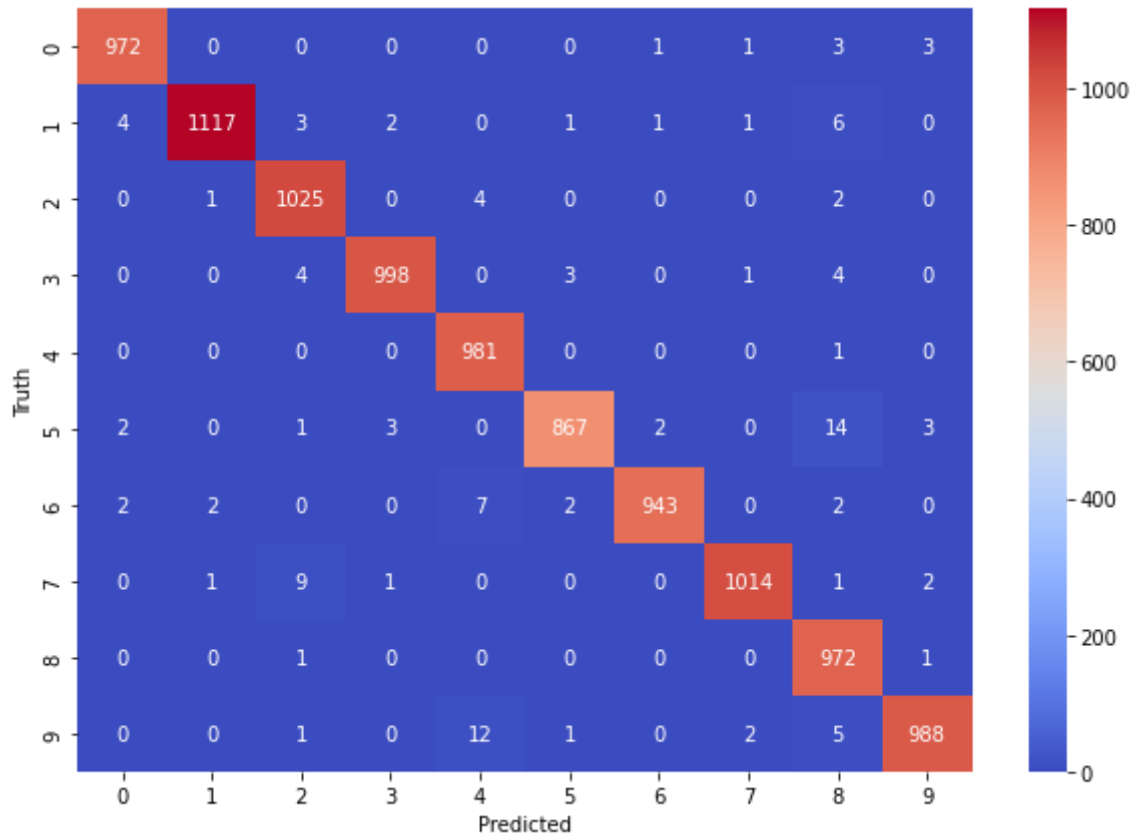
```
# Confusion Matrix
y_predicted = model.predict(x_test)
y_predicted_labels = [np.argmax(i) for i in y_predicted]
cm = tf.math.confusion_matrix(labels=y_test, predictions=y_predicted_labels)
cm
```

```
<tf.Tensor: shape=(10, 10), dtype=int32, numpy=
array([[ 972,    0,    0,    0,    0,    0,    1,    1,    3,    3],
       [   4, 1117,    3,    2,    0,    1,    1,    1,    6,    0],
       [   0,    1, 1025,    0,    4,    0,    0,    0,    2,    0],
       [   0,    0,    4,  998,    0,    3,    0,    1,    4,    0],
       [   0,    0,    0,    0,  981,    0,    0,    0,    1,    0],
       [   2,    0,    1,    3,    0,  867,    2,    0,   14,    3],
       [   2,    2,    0,    0,    7,    2,  943,    0,    2,    0],
       [   0,    1,    9,    1,    0,    0,    0, 1014,    1,    2],
       [   0,    0,    1,    0,    0,    0,    0,    0,  972,    1],
       [   0,    0,    1,    0,   12,    1,    0,    2,    5,  988]])>
```

▼ Plotting heat map

```
plt.figure(figsize = (10,7))
sns.heatmap(cm, annot=True, fmt='d', cmap = 'coolwarm')
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(69.0, 0.5, 'Truth')



▼ Testing the Model

```
# Testing the Model
plt.imshow(x_test[7], cmap='gray_r')
plt.title('Actual Value: {}'.format(y_test[7]))
prediction=model.predict(x_test)

plt.axis('off')
print('Predicted Value: ', np.argmax(prediction[7]))
if(y_test[7]==(np.argmax(prediction[7]))):
    print('Successful prediction')
else:
    print('Unsuccessful prediction')
```

Predicted Value: 9
Successful prediction

```
plt.imshow(x_test[1],cmap='gray_r')  
plt.title('Actual Value: {}'.format(y_test[1]))  
prediction=model.predict(x_test)  
plt.axis('off')  
print('Predicted Value: ',np.argmax(prediction[1]))  
if(y_test[1]==(np.argmax(prediction[1]))):  
    print('Successful prediction')  
else:  
    print('Unsuccessful prediction')
```



Predicted Value: 2
Successful prediction
Actual Value: 2

