import library

In [1]:

```
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers, models
import numpy as np
import matplotlib.pyplot as plt
```

Get MNIST Data.

MNIST data loacted in tensorflow > keras > datasets > mnist

Split data to (train images, train labels) and (test images, test labels)

```
In [2]:
```

```
mnist = keras.datasets.mnist
(train_images, train_labels), (test_images, test_labels) = mnist.load_data()
```

There are Total 60000 Train images and Train labels. (6000 images for single class)

Shape of single image is 28 x 28 (pixel)

```
In [3]:
```

```
print('Shape of Train images :',train_images.shape)
print('Shape of Train labels : ', train_labels.shape)
print('WnShape of Test images : ', test_images.shape)
print("Shape of Test labels : ",test_labels.shape)

Shape of Train images : (60000, 28, 28)
Shape of Train labels : (60000,)

Shape of Test images : (10000, 28, 28)
Shape of Test labels : (10000,)

In [4]:

print('Train labels : ',train_labels)
```

Plot first train image.

Train labels: [5 0 4 ... 5 6 8]

when value is close to 0 : dark

when value is close to 255: white

In [5]:

Print(train_images[1])														_
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	<pre>print(train_images[1])</pre>													
							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	
159 50 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	51	159	253	
C	$[\ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ $				0	0	0	0	0	48	238	252	252	
[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	_						0	0	54	227	253	252	239	
84 252 253 122 0	233 252 57 6 0 0	0	0	0	0]									
96 189 253 167 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	84 252 253 122 0 0	0	0	0	0]									
47 79 255 168 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	96 189 253 167 0 0	0	0	0	0]									
0 0 253 243 50 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	47 79 255 168 0 0	0	0	0	0]									
0 0 253 252 165 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 253 243 50 0	0	0	0	0]									
0 0 253 252 195 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 0 253 252 195 0 0 0 0 0 0 0] [0 0 0 253 252 195 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 0 253 252 165 0	0	0	0	0]									
0 0 253 252 195 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 253 252 195 0	0	0	0	0]									
[0	-						0	0	0	0	0	0	0	
[0 0 0 0 0 0 0 0 0 76 246 252 112 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 85 252 230 25 0 0 0 0 0 0 0 0 0 0 0 0 0 7 135 253 186 12 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	[0 0 0 0 0 0	76	246	252	112	0	0	0	0	0	0	0	0	
[0 0 0 0 0 0 0 0 0 85 252 223 0 0 0 0 0 0 0 0 0 0 0 0 0 7 131 252 225 71 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]	85	252	230	25	0	0	0	0	0	0	0	0	
[0 0 0 0 0 0 0 0 85 252 145 0 0 0 0 0 0 0 0 48 165 252 173 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	85	252	223	0	0	0	0	0	0	0	0	7	
[0 0 0 0 0 0 0 0 86 253 225 0 0 0 0 0 0 114 238 253 167 62 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 114 238 253 167 60 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	[0 0 0 0 0 0	85	252	145	0	0	0	0	0	0	0	48	165	
[0 0 0 0 0 0 0 85 252 249 146 48 29 85 178 225 253 223 167 56 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	[0 0 0 0 0 0	86	253	225	0	0	0	0	0	0	114	238	253	
[0 0 0 0 0 0 0 85 252 252 252 252 229 215 252 252 252 196 130 0 0 0 0 0 0 0 0 0 0 0] [0 0 0 0 0 0 0 0 0 28 199 252 252 253 252 252 233 145 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	85	252	249	146	48	29	85	178	225	253	223	167	
[0 0 0 0 0 0 0 28 199 252 252 253 252 252 233 145 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			252	252	252	229	215	252	252	252	196	130	0	
[0 0 0 0 0 0 0 0 0 25 128 252 253 252 141 37 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	_						252	252	233	145	0	0	0	
0 0							252	141	37	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]									
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	0 0 0 0 0 0	0	0	0	0]									
	0 0 0 0 0 0	0	0	0	0]									
	_						U	U	U	U	U	U	U	

Plot First 10 Train images and Corresponding labels

In [6]:

```
print('First 10 Train images in MNIST dataset\n')
for i in range(10):
    plt.subplot(1, 10, i+1)
    plt.xticks([])
    plt.yticks([])
    plt.imshow(train_images[i])
plt.show()
print('\nTrain labels match with Train label sequentialy\n',train_labels[:10])
```

First 10 Train images in MNIST dataset



Train labels match with Train label sequentialy [5 0 4 1 9 2 1 3 1 4]

Important

Change data shape (60000 x 28 x 28) to (60000 x 28 x 28 x 1)

```
In [7]:
```

```
train_images = tf.reshape(train_images, [-1, 28, 28, 1])
test_images = tf.reshape(test_images, [-1, 28, 28, 1])
```

Select one convolution model below

There are 3 example models.

3, 5, 7 layer each

MODEL 1: 3 Layers with 1 Convolution layer

MODEL 2 : 5 Layers with 2 Convolution layer

MODEL 3:7 Layers with 4 Convolution layer

In [8]:

```
def select_model(model_number):
    if model_number == 1:
       model = keras.models.Sequential([
                    keras.layers.Conv2D(32, (3,3), activation = 'relu', input_shape = (28, 28,1)),
                   keras.layers.MaxPool2D((2,2)),
                   keras.layers.Flatten(),
                    keras.layers.Dense(10, activation = 'softmax')])
    if model_number == 2:
       model = keras.models.Sequential([
                    keras.layers.Conv2D(32, (3,3), activation = 'relu', input_shape=(28,28,1)),
                   keras.layers.MaxPool2D((2,2)),
                   keras.layers.Conv2D(64, (3,3), activation = 'relu'),
                   keras.layers.MaxPool2D((2,2)),
                    keras.layers.Flatten(),
                    keras.layers.Dense(10, activation = 'softmax')])
    if model_number == 3:
       model = keras.models.Sequential([
                    keras.layers.Conv2D(32, (3,3), activation = 'relu', input_shape = (28, 28,1)),
                   keras.layers.MaxPool2D((2,2)),
                    keras.layers.Conv2D(64, (3,3), activation = 'relu'),
                    keras.layers.Conv2D(64, (3.3), activation = 'relu').
                    keras.layers.MaxPool2D((2,2)),
                    keras.layers.Conv2D(128, (3.3), activation = 'relu'),
                    keras.layers.Flatten(),
                    keras.layers.Dense(10, activation = 'softmax')])
   return model
```

```
In [9]:
```

```
model = select_model(1)
```

If you want to see information of model, model.summary() will help

summary() is also built in function

In [10]:

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 13, 13, 32)	0
flatten (Flatten)	(None, 5408)	0
dense (Dense)	(None, 10)	54090

Total params: 54,410 Trainable params: 54,410 Non-trainable params: 0

Components in training step

Optimizer, Loss function, accuracy metrics

```
In [11]:
```

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

Training Step

Training for 5 epochs.

```
In [12]:
```

```
model.fit(train_images, train_labels, epochs = 5)
Epoch 1/5
                           ========] - 18s 9ms/step - loss: 0.6037 - accuracy:
1875/1875 [=====
0.9406
Epoch 2/5
                       ===============] - 21s 11ms/step - loss: 0.0857 - accurac
1875/1875 [======
y: 0.9752
Epoch 3/5
                             =======] - 20s 11ms/step - loss: 0.0685 - accurac
1875/1875 [===
v: 0.9793
Epoch 4/5
1875/1875 [=====
                           ========] - 19s 10ms/step - loss: 0.0618 - accurac
y: 0.9815
Epoch 5/5
1875/1875 [======
                       ========== ] - 19s 10ms/step - loss: 0.0489 - accurac
y: 0.9848
Out[12]:
<keras.callbacks.History at 0x1a5f0da7e20>
```

Test Step

Perform Test with Test data

```
In [13]:
```

```
test_loss, accuracy = model.evaluate(test_images, test_labels, verbose = 2)

print('\text_loss : ', test_loss)

print('Test_accuracy : ', accuracy)

313/313 - 1s - loss: 0.1184 - accuracy: 0.9744 - 1s/epoch - 3ms/step

Test_loss : 0.11840185523033142

Test_accuracy : 0.974399983882904
```

Before prediction, change test image's type to float 32.

```
In [14]:
```

Prediction: (10000, 10) Test labels: (10000,)

Functions for plot images, probability

In [16]:

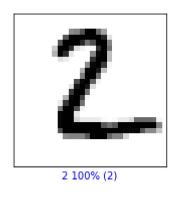
```
def plot_image(i, predictions_array, true_label, img):
  predictions_array, true_label, img = predictions_array[i], true_label[i], img[i]
  plt.grid(False)
  plt.xticks([])
  plt.yticks([])
  plt.imshow(img, cmap=plt.cm.binary)
  predicted_label = np.argmax(predictions_array)
  if predicted_label == true_label:
    color = 'blue
  else:
    color = 'red'
  plt.xlabel("{} {:2.0f}% ({})".format(Number[predicted_label],
                                100*np.max(predictions_array),
                                Number[true_label]),
                                color=color)
def plot_value_array(i, predictions_array, true_label):
  predictions_array, true_label = predictions_array[i], true_label[i]
  plt.grid(False)
  plt.xticks([])
  plt.yticks([])
  thisplot = plt.bar(range(10), predictions_array, color="#777777")
  plt.ylim([0, 1])
  predicted_label = np.argmax(predictions_array)
  plt.xticks(Number)
  thisplot[predicted label].set color('red')
  thisplot[true_label].set_color('blue')
```

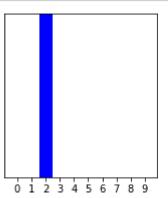
In [17]:

```
(train_images, train_labels), (test_images, test_labels) = mnist.load_data()
```

In [18]:

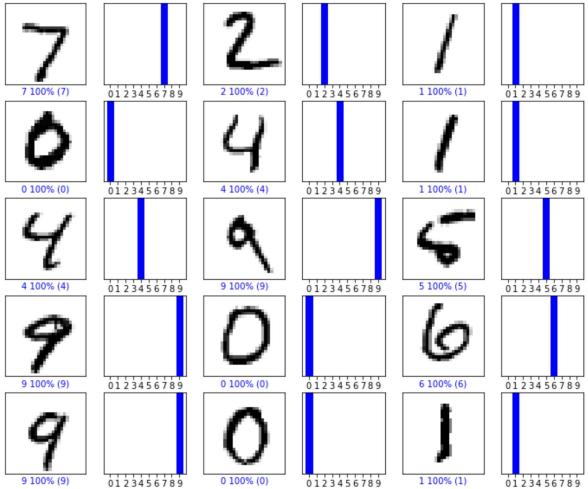
```
i = 1
plt.figure(figsize=(6,3))
plt.subplot(1,2,1)
plot_image(i, pred, test_labels, test_images)
plt.subplot(1,2,2)
plot_value_array(i, pred, test_labels)
plt.show()
```





In [19]:

```
num_rows = 5
num_cols = 3
num_images = num_rows*num_cols
plt.figure(figsize=(2*2*num_cols, 2*num_rows))
for i in range(num_images):
    plt.subplot(num_rows, 2*num_cols, 2*i+1)
    plot_image(i, pred, test_labels, test_images)
    plt.subplot(num_rows, 2*num_cols, 2*i+2)
    plot_value_array(i, pred, test_labels)
plt.show()
```



Plot images and probability that model predicted wrong

In [20]:

```
def error_mnist(prediction_array, true_label):
    error_index = []
    for i in range(true_label.shape[0]):
        if np.argmax(prediction_array[i]) != true_label[i]:
            error_index.append(i)
    return error_index
# change num_cols, num_rows if you want to see more result.
def plot_error(index, prediction_array, true_label):
    num\_cols = 5
    num_rows = 5
    plt.figure(figsize=(2*2*num_cols, 2*num_rows))
    assert len(index) < num_cols * num_rows</pre>
    for i in range(len(index)):
        plt.subplot(num_rows, 2*num_cols, 2*i+1)
        idx = index[i]
        plt.imshow(test_images[idx])
        plt.subplot(num_rows, 2*num_cols, 2*i+2)
        plt.bar(range(10), prediction_array[idx])
        plt.xticks(Number)
```

Find index of wrong prediction

Plot first 10 wrong predicted images and probability

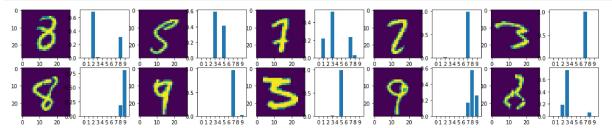
In [21]:

```
index = error_mnist(pred, test_labels)
index_slice = index[:10]
print(index[:10])
```

[184, 211, 282, 321, 381, 403, 417, 449, 488, 582]

In [22]:

```
plot_error(index_slice, pred, test_labels)
```



DONE

NameError
Input In [23], in <cell line: 1>()
----> 1 DONE

NameError: name 'DONE' is not defined

In []: