

```
#CNN is a type of artificial Neural Network which is widely used for object/image recognition and classification.  
#Deep Learning recognizes objects in an image by using CNN.
```



```
# Import the necessary packages  
import tensorflow as tf  
from keras.models import Sequential  
from keras.layers import Dense,Conv2D,Dropout,Flatten,MaxPooling2D  
import matplotlib.pyplot as plt  
import numpy as np  
  
# a. Loading and preprocessing the image data  
mnist = tf.keras.datasets.mnist  
(x_train, y_train), (x_test, y_test) = mnist.load_data()  
input_shape = (28,28,1) # images are greyscale thats why input channel is 1  
  
# making sure that the values are float so that we can get the decimal points after devision  
x_train = x_train.reshape(x_train.shape[0],28,28,1)  
x_test = x_test.reshape(x_test.shape[0],28,28,1)  
  
# print('Data type of x_train:',x_train.dtype)  
  
x_train = x_train.astype('float32')  
x_test = x_test.astype('float32')  
  
# print('Data type of x_train after converting to float:',x_train.dtype)
```

```
# Normalizing the RGB codes by dividing it to the max RGB value  
x_train = x_train/255  
x_test = x_test/255  
print('shape of training :',x_train.shape)
```

```
shape of training : (60000, 28, 28, 1)  
  
print('shape of testing :',x_test.shape)
```

```
shape of testing : (10000, 28, 28, 1)
```

```
# b. Defining the model's architecture  
model = Sequential() # used sequential as we have to add layers one after another  
model.add(Conv2D(28, kernel_size=(3,3),input_shape=input_shape))  
# kernel size - it is kernel size or filter size, it is an size of each convolutional layer, you can change size also  
# input shape is input size which we have declared above  
model.add(MaxPooling2D(pool_size=(2,2)))  
model.add(Flatten())  
model.add(Dense(200,activation='relu')) # Hidden Layer  
model.add(Dropout(0.3)) # Will drop some random neurons from hidden layer, 30%neurons will be removed  
model.add(Dense(10,activation='softmax'))  
model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 26, 26, 28)	280
=====		
max_pooling2d (MaxPooling2D)	(None, 13, 13, 28)	0
flatten (Flatten)	(None, 4732)	0

```

dense (Dense)           (None, 200)      946600
dropout (Dropout)       (None, 200)      0
dense_1 (Dense)         (None, 10)       2010
=====
# c. Training the model

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

model.fit(x_train,y_train,epochs=2)

Epoch 1/2
1875/1875 [=====] - 70s 36ms/step - loss: 0.2018 - accuracy: 0.9387
Epoch 2/2
1875/1875 [=====] - 67s 36ms/step - loss: 0.0833 - accuracy: 0.9746
<keras.src.callbacks.History at 0x22a01f50c10>

```

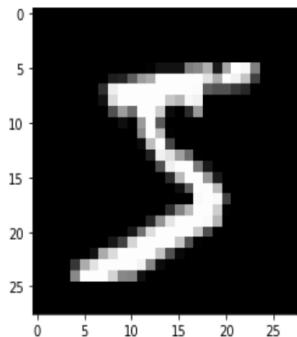
```

# d. Estimating the model's performance

test_loss, test_acc = model.evaluate(x_test, y_test)
print('loss=% .3f' %test_loss)
print('Accuracy=% .3f' %test_acc)

313/313 [=====] - 4s 11ms/step - loss: 0.0654 - accuracy: 0.9786
loss=0.065
Accuracy=0.979
# Showing image at position[] from dataset
image = x_train[0]
plt.imshow(np.squeeze(image),cmap='gray') # it will remove arrays of single dimensional
plt.show()

```



```

# predicting the class of image
image = image.reshape(1,image.shape[0],image.shape[1],image.shape[2])
predict_model = model.predict([image])
print('predicted class: {}'.format(np.argmax(predict_model))) # it displays max value

1/1 [=====] - 0s 216ms/step
predicted class: 5

```