

#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 chaneel 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 deviding 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

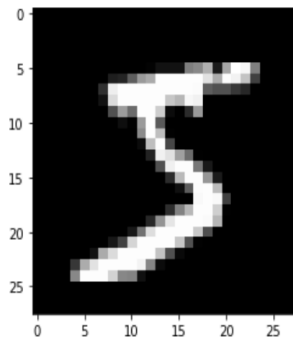
Accuracy=0.979

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