

```

import tensorflow
from tensorflow import keras
from tensorflow.keras import Sequential
from tensorflow.keras.layers import Dense , Flatten

(x_train , y_train ) , (x_test , y_test) = keras.datasets.mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 0s 0us/step

x_train.shape

(60000, 28, 28)

len(x_train)

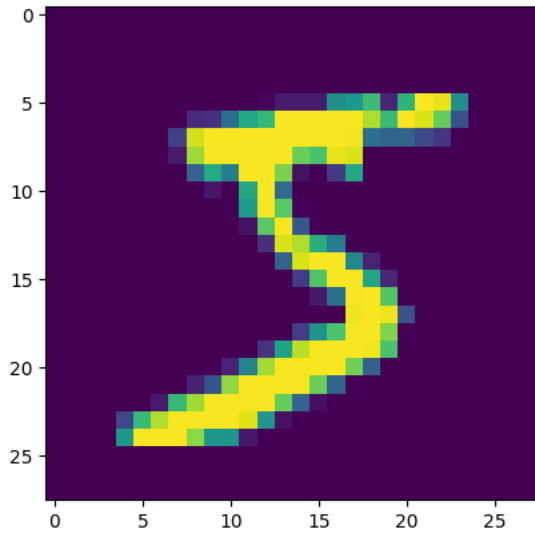
60000

x_train[1]
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 51, 238, 253,
253, 190, 114, 253, 228, 47, 79, 255, 168, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 48, 238, 252, 252,
179, 12, 75, 121, 21, 0, 0, 253, 243, 50, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 38, 165, 253, 233, 208,
84, 0, 0, 0, 0, 0, 0, 253, 252, 165, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 7, 178, 252, 240, 71, 19,
28, 0, 0, 0, 0, 0, 0, 253, 252, 195, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 57, 252, 252, 63, 0, 0,
0, 0, 0, 0, 0, 0, 253, 252, 195, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 198, 253, 190, 0, 0, 0,
0, 0, 0, 0, 0, 0, 255, 253, 196, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 76, 246, 252, 112, 0, 0, 0,
0, 0, 0, 0, 0, 0, 253, 252, 148, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 85, 252, 230, 25, 0, 0, 0,
0, 0, 0, 0, 7, 135, 253, 186, 12, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 85, 252, 223, 0, 0, 0, 0,
0, 0, 0, 0, 7, 131, 252, 225, 71, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 85, 252, 145, 0, 0, 0, 0,
0, 0, 0, 48, 165, 252, 173, 0, 0, 0, 0,
0, 0],
[ 0, 0, 0, 0, 0, 0, 86, 253, 225, 0, 0, 0, 0,
0, 0, 114, 238, 253, 162, 0, 0, 0, 0,
0, 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, 85, 252, 252, 252, 229, 215, 252,
252, 252, 196, 130, 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, 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, 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, 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]], dtype=uint8)

```

```
import matplotlib.pyplot as plt  
plt.imshow(x_train[0])
```

<matplotlib.image.AxesImage at 0x7f20b466edd0>



```
x_train = x_train / 255.  
x_test = x_test / 255.
```

```
x_train[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.      , ]])

```

```
model = Sequential()
```

```

model.add(Flatten(input_shape = (28 , 28)))
model.add(Dense(128 , activation = "relu"))
model.add(Dense(10 ,activation = 'softmax'))

```

```
model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
=====		
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 128)	100480
dense_1 (Dense)	(None, 10)	1290
=====		
Total params: 101,770		
Trainable params: 101,770		
Non-trainable params: 0		

```
model.compile(loss = "sparse_categorical_crossentropy" , optimizer = "Adam")
```

```
model.fit(x_train , y_train , epochs = 15 , validation_split = 0.2)
```

```

Epoch 1/15
1500/1500 [=====] - 8s 5ms/step - loss: 0.2848 - val_loss: 0.1588
Epoch 2/15
1500/1500 [=====] - 5s 4ms/step - loss: 0.1254 - val_loss: 0.1101
Epoch 3/15
1500/1500 [=====] - 7s 5ms/step - loss: 0.0864 - val_loss: 0.0999
Epoch 4/15
1500/1500 [=====] - 6s 4ms/step - loss: 0.0651 - val_loss: 0.0888
Epoch 5/15
1500/1500 [=====] - 7s 5ms/step - loss: 0.0493 - val_loss: 0.0875
Epoch 6/15
1500/1500 [=====] - 6s 4ms/step - loss: 0.0402 - val_loss: 0.0867
Epoch 7/15
1500/1500 [=====] - 7s 4ms/step - loss: 0.0299 - val_loss: 0.0961
Epoch 8/15
1500/1500 [=====] - 6s 4ms/step - loss: 0.0252 - val_loss: 0.0941
Epoch 9/15
1500/1500 [=====] - 7s 4ms/step - loss: 0.0189 - val_loss: 0.0938
Epoch 10/15
1500/1500 [=====] - 6s 4ms/step - loss: 0.0173 - val_loss: 0.1040
Epoch 11/15
1500/1500 [=====] - 7s 4ms/step - loss: 0.0154 - val_loss: 0.0930
Epoch 12/15
1500/1500 [=====] - 5s 4ms/step - loss: 0.0104 - val_loss: 0.0983
Epoch 13/15
1500/1500 [=====] - 12s 8ms/step - loss: 0.0119 - val_loss: 0.1027
Epoch 14/15
1500/1500 [=====] - 9s 6ms/step - loss: 0.0085 - val_loss: 0.1116
Epoch 15/15
1500/1500 [=====] - 6s 4ms/step - loss: 0.0075 - val_loss: 0.1158
<keras.callbacks.History at 0x7f20b5ec2cb0>

```

```

y_prob = model.predict(x_test)
y_prob

```

```

313/313 [=====] - 0s 2ms/step
array([[6.6217028e-12, 4.9329313e-10, 8.7182173e-09, ..., 9.9996823e-01,

```

```

5.2964433e-10, 2.0152461e-06],
[2.7174196e-16, 2.4983011e-08, 9.9999994e-01, ..., 8.1326343e-25,
 2.1685561e-14, 3.2354656e-27],
[3.2452921e-10, 9.9996811e-01, 6.6366424e-06, ..., 2.6902881e-06,
 2.1086196e-05, 4.5902857e-10],
...,
[2.7549790e-20, 7.7689950e-14, 2.7030318e-21, ..., 8.5932268e-11,
 3.6087779e-11, 2.4283278e-08],
[7.4249279e-18, 5.5016094e-21, 1.3112478e-18, ..., 1.6592871e-18,
 5.4723159e-10, 3.1802546e-19],
[5.2764455e-17, 1.7015963e-22, 3.2704603e-11, ..., 1.3154449e-20,
 8.9802969e-17, 2.3722108e-16]], dtype=float32)

```

```
y_pred = y_prob.argmax(axis = 1)
```

```
from sklearn.metrics import accuracy_score
```

```
accuracy_score(y_test , y_pred)
```

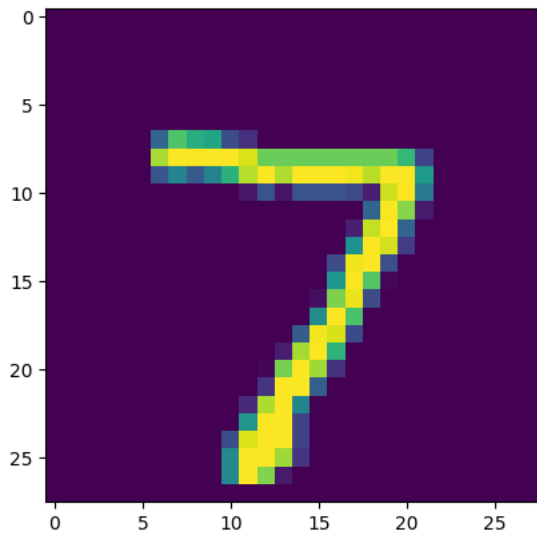
```
0.9767
```

```

# checking the model
plt.imshow(x_test[0])

```

```
<matplotlib.image.AxesImage at 0x7f2083f81b10>
```



```
model.predict(x_test[0].reshape(1 , 28 , 28 )).argmax(axis = 1)
```

```

1/1 [=====] - 0s 36ms/step
array([7])

```

```

# checking the model
plt.imshow(x_test[1])

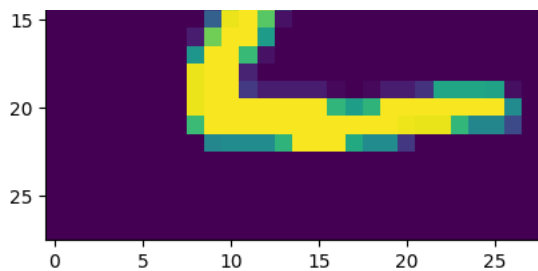
```

```
<matplotlib.image.AxesImage at 0x7f2083fd47f0>
```



```
model.predict(x_test[1].reshape(1 , 28 , 28)).argmax(axis = 1)
```

```
1/1 [=====] - 0s 41ms/step  
array([2])
```



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