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In [2]: import numpy as np
import tensorflow as tf    #https://www.youtube.com/watch?v=\_c\_x8A3mNDk
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In [3]: from tensorflow import keras
import matplotlib.pyplot as plt
```

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In [5]: data=tf.keras.datasets.mnist
from tensorflow.keras.optimizers import SGD
from tensorflow.keras.datasets import mnist
```

```
In [6]: (x_train,y_train),(x_test,y_test)=data.load_data()    # x is attribute and y
```

```
In [7]: x_test.shape
```

Out[7]: (10000, 28, 28)

```
In [23]: x_train.shape    # show the data set of training    28 pixel , 28 pixel pt
print(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.	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.

```
In [7]: # To perform Machine Learning, it is important to convert all the values fr
# The simplest way is to divide the value of every pixel by 255 to get the
```

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In [8]: x_train,x_test =x_train/255,x_test/255      # Covert all values between 0 to 1
```

```
In [9]: model=tf.keras.models.Sequential([
tf.keras.layers.Flatten(input_shape=(28,28)),      # covert 2D into 1D
tf.keras.layers.Dense(150,activation='relu'),      # Dense Means Fully con
tf.keras.layers.Dense(10,activation='softmax')     # classifying into 10 cl
])
```

```
In [10]: # The default Learning rate is 0.01 and no momentum is used by default.
sgd=SGD(0.02)   # is Learning rate 0.02
#adam
model.compile(optimizer='sgd', loss='sparse_categorical_crossentropy', metr
#
```

```
In [11]: history=model.fit(x_train, y_train,validation_data=(x_test,y_test),epochs=5
```

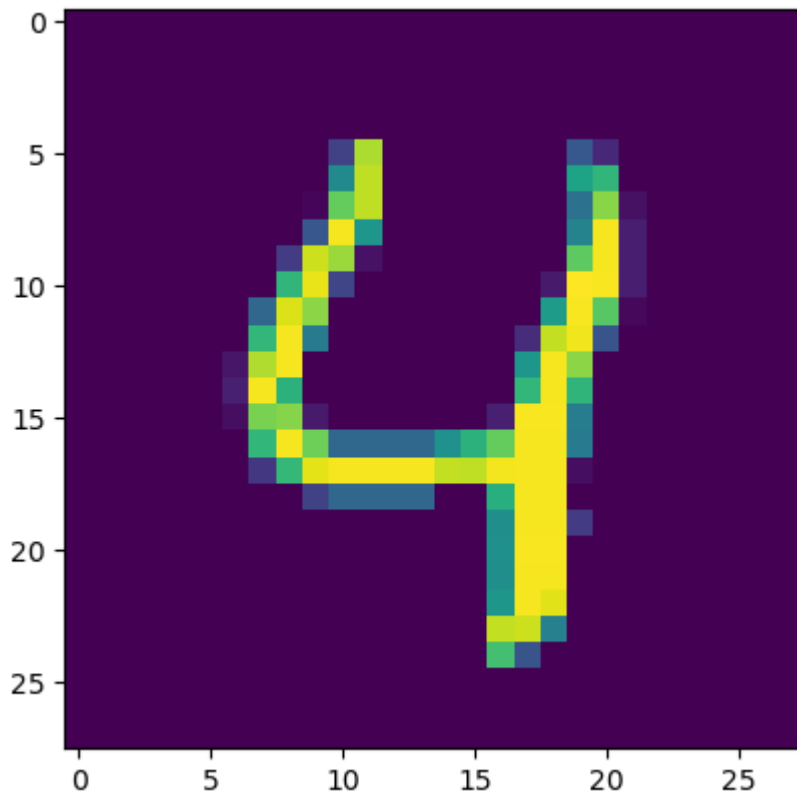
```
Epoch 1/5
1875/1875 [=====] - 10s 4ms/step - loss: 0.6422
- accuracy: 0.8350 - val_loss: 0.3570 - val_accuracy: 0.9026
Epoch 2/5
1875/1875 [=====] - 5s 3ms/step - loss: 0.3351 -
accuracy: 0.9075 - val_loss: 0.2902 - val_accuracy: 0.9184
Epoch 3/5
1875/1875 [=====] - 6s 3ms/step - loss: 0.2849 -
accuracy: 0.9209 - val_loss: 0.2585 - val_accuracy: 0.9279
Epoch 4/5
1875/1875 [=====] - 6s 3ms/step - loss: 0.2544 -
accuracy: 0.9289 - val_loss: 0.2368 - val_accuracy: 0.9347
Epoch 5/5
1875/1875 [=====] - 6s 3ms/step - loss: 0.2314 -
accuracy: 0.9358 - val_loss: 0.2165 - val_accuracy: 0.9400
```

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In [14]: # Evaluate the model
test_loss,test_acc=model.evaluate(x_test,y_test)
```

```
313/313 [=====] - 1s 2ms/step - loss: 0.2179 - a
ccuracy: 0.9383
```

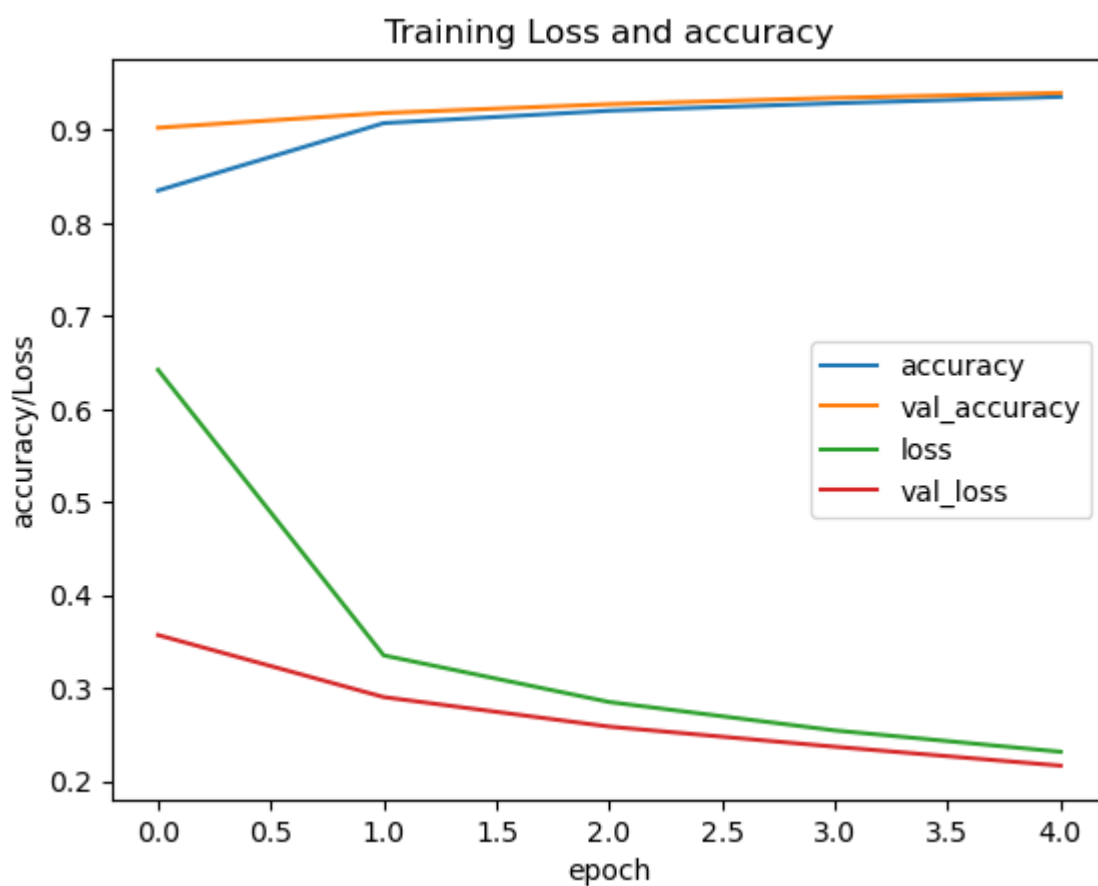
```
In [12]: plt.imshow(x_test[4])  
prediction=model.predict(x_test) #predict the data  
print(np.argmax(prediction[4])) # print data depend on max probaiblites  
#plt.show()
```

313/313 [=====] - 1s 2ms/step
4



```
In [14]: # graph represents the model's loss
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])

plt.title('Training Loss and accuracy')
plt.ylabel('accuracy/Loss')
plt.xlabel('epoch')
plt.legend(['accuracy', 'val_accuracy', 'loss', 'val_loss'])
plt.show()
```



In []: