111 [ ].	## Name: Satyam kumar ## Roll no: I4250 ## Subject:LP-IV(DL)
In [ ]: In [1]:	!pip install tensorflow  #importing necessary libraries
	<pre>import tensorflow as tf from tensorflow import keras import pandas as pd import numpy as np</pre>
	<pre>import matplotlib.pyplot as plt import random %matplotlib inline</pre>
In [2]:	<pre>#import dataset and split into train and test data mnist = tf.keras.datasets.mnist (x_train, y_train), (x_test, y_test) = mnist.load_data()</pre>
<pre>In [3]: Out[3]:</pre>	<pre>plt.matshow(x_train[1]) <matplotlib.image.axesimage 0x216d2a9c9d0="" at=""></matplotlib.image.axesimage></pre>
	0 5 10 15 20 25
	10 -
	20 -
	25 -
In [4]: Out[4]:	<pre>plt.imshow(-x_train[0], cmap="gray") <matplotlib.image.axesimage 0x216d321cac0="" at=""></matplotlib.image.axesimage></pre>
	5 -
	20 - 25 -
In [5]:	0 5 10 15 20 25 x_train = x_train / 255
In [6]:	<pre>x_test = x_test / 255  model = keras.Sequential([</pre>
	<pre>keras.layers.Flatten(input_shape=(28, 28)), keras.layers.Dense(128, activation="relu"), keras.layers.Dense(10, activation="softmax") ])</pre>
	<pre>model.summary()  Model: "sequential"</pre>
	Layer (type)
	dense_1 (Dense) (None, 10) 1290  ===================================
	Trainable params: 101,770 Non-trainable params: 0
111 [7].	<pre>model.compile(optimizer="sgd", loss="sparse_categorical_crossentropy", metrics=['accuracy'])</pre>
In [8]:	<pre>history=model.fit(x_train, y_train,validation_data=(x_test,y_test),epochs=10)</pre> Epoch 1/10
	1875/1875 [====================================
	1875/1875 [====================================
	Epoch 6/10  1875/1875 [====================================
	1875/1875 [====================================
In [9]:	<pre>test_loss, test_acc=model.evaluate(x_test, y_test) print("Loss=%.3f" %test_loss) print("Accuracy=%.3f" %test_acc)</pre>
	313/313 [===================================
In [10]:	<pre>n=random.randint(0,9999) plt.imshow(x_test[n]) plt.show()</pre>
	5 -
	10 -
	20 -
In [11]:	0 5 10 15 20 25 x_train
Out[11]:	array([[[0., 0., 0.,, 0., 0., 0.],
	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]],
	[[0., 0., 0.,, 0., 0., 0.], [0., 0., 0.,, 0., 0., 0.], [0., 0., 0.,, 0., 0.], , [0., 0., 0.,, 0., 0.],
	[0., 0., 0.,, 0., 0., 0.], [0., 0., 0.,, 0., 0.], [[0., 0., 0.,, 0., 0.], [[0., 0., 0.,, 0., 0.],
	[o., o., o., o., o., o., o.],, [o., o., o.,, o., o.],
	[[0., 0., 0.,, 0., 0.], [[0., 0., 0.,, 0., 0.],
	[0., 0., 0.,, 0., 0., 0.], , [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],
	[0., 0., 0.,, 0., 0., 0.]], [[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],
	, [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 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 [12]:	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]]) x_test
Out[12]:	array([[[0., 0., 0.,, 0., 0., 0.],
	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]],
	[[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], , [0., 0., 0.,, 0., 0.],
	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]], [[0., 0., 0.,, 0., 0.]], [[0., 0., 0.,, 0., 0.],
	[0., 0., 0., 0., 0., 0.],, [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],
	···, [[0., 0., 0.,, 0., 0.], [0., 0., 0., 0., 0.],
	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],, [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],
	[[0., 0., 0.,, 0., 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.],
	, [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]], [0., 0., 0., 0.]
	[[0., 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 [13]:	[0., 0., 0.,, 0., 0.], [0., 0., 0.,, 0., 0.]]])  predicted_value=model.predict(x_test) plt.imshow(x_test[n])
	<pre>plt.show() print(predicted_value[n])</pre>
	313/313 [===================================
	10 -
	20 -
	5 10 15 20 25 [5.5406590e-06 9.9045217e-01 9.6445804e-04 1.1881288e-03 5.5083026e-05 1.0767796e-03 5.1782565e-04 5.0224096e-04 4.8863539e-03 3.5130000e-04]
In [14]:	<pre># history.history() history.history.keys() # dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])</pre>
	<pre>plt.plot(history.history['accuracy']) plt.plot(history.history['val_accuracy']) plt.title('model accuracy') plt.ylabel('accuracy')</pre>
	<pre>plt.ylabel('accuracy') plt.xlabel('epoch') plt.legend(['Train', 'Validation'], loc='upper left') plt.show()</pre>
	model accuracy  Train Validation
	0.92 - \$\frac{1}{2} 0.90 - \frac{1}{2} 0.88 - \fra
	0.86
	0.84 - /
In [15]:	<pre># history.history() history.history.keys() # dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])</pre>
	<pre>plt.plot(history.history['loss']) plt.plot(history.history['val_loss']) plt.title('model loss') plt.ylabel('loss')</pre>
	<pre>plt.ylabel('loss') plt.xlabel('epoch') plt.legend(['Train', 'Validation'], loc='upper left') plt.show()</pre>
	model loss  Train Validation
	0.5 - <u>§</u> 0.4 -
	0.3
	0.2 1
In [ ]:	