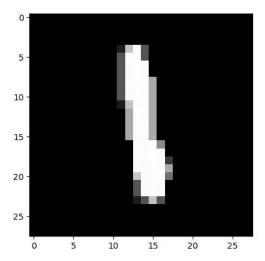
```
In [ ]: Shirsath Vaishnavi
          Roll No: 61
In [100]: !pip install keras
          Requirement already satisfied: keras in c:\users\praja\anaconda3\lib\site-packages (2.10.0)
In [101]: import numpy as np
          from keras.datasets import mnist
          from keras.models import Sequential
          from keras.layers import Dense
In [102]: (x_train, y_train),(x_test, y_test) = mnist.load_data()
In [103]: x_train.shape
Out[103]: (60000, 28, 28)
In [104]: x_test.shape
Out[104]: (10000, 28, 28)
In [105]: import matplotlib.pyplot as plt
In [106]: x = np.zeros(100).reshape(10,10)
In [107]: plt.imshow(x, cmap='gray')
Out[107]: <matplotlib.image.AxesImage at 0x205a11d0430>
           0
           2
           4
           6
           8
                                                      8
In [108]: plt.imshow(x_train[200], cmap='gray')
```

Out[108]: <matplotlib.image.AxesImage at 0x205a1204a30>

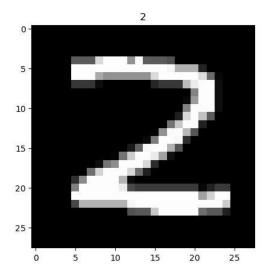


In [109]: y_train[200]

Out[109]: 1

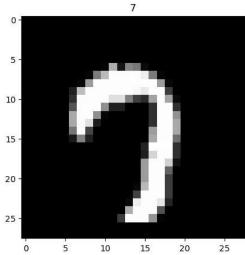
```
In [110]: plt.imshow(x_train[220], cmap='gray')
plt.title(y_train[220])
```

```
Out[110]: Text(0.5, 1.0, '2')
```



```
In [111]: plt.imshow(x_test[220], cmap='gray')
plt.title(y_test[220])
```

```
Out[111]: Text(0.5, 1.0, '7')
```



```
In [112]: x = np.array([[2,3,5],[8,9,0]])
In [113]: x
Out[113]: array([[2,3,5],[8,9,0]])
In [114]: x.shape
Out[114]: (2, 3)
In [115]: x = x.flatten()
x
Out[115]: array([2,3,5,8,9,0])
In [116]: x.shape
Out[116]: (6,)
In [117]: img = x_train[3]
In [118]: img.shape
Out[118]: (28, 28)
```

```
In [119]: | img = img.flatten()
          img.shape
Out[119]: (784,)
In [120]: x_train = x_train.reshape(60000, 784)
           x_{\text{test}} = x_{\text{test.reshape}}(10000, 784)
In [121]: x_train.shape
Out[121]: (60000, 784)
In [122]: x = \text{np.array}([8,6,5,7,0,3,4,2])
In [123]: x/8
Out[123]: array([1. , 0.75 , 0.625, 0.875, 0. , 0.375, 0.5 , 0.25 ])
In [124]: x_train = x_train / 255
In [125]: x_test = x_test / 255
In [126]: set(y_train)
Out[126]: {0, 1, 2, 3, 4, 5, 6, 7, 8, 9}
In [127]: import seaborn as sns
In [128]: sns.countplot(x = y_train)
Out[128]: <AxesSubplot:ylabel='count'>
              7000
              6000
              5000
               4000
              3000
              2000
              1000
                  0
                                                         5
                       0
                              1
In [129]: from collections import Counter
          Counter(y_train)
Out[129]: Counter({5: 5421,
                    0: 5923,
                    4: 5842,
                    1: 6742,
                    9: 5949,
                    2: 5958,
                    3: 6131,
                    6: 5918,
                    7: 6265,
                    8: 5851)
In [130]: from keras.utils import to_categorical
In [131]: x = [0,2,2,1,0,1,2]
In [132]: to_categorical(x)
Out[132]: array([[1., 0., 0.],
                  [0., 0., 1.],
                  [0., 0., 1.],
[0., 1., 0.],
                  [1., 0., 0.],
[0., 1., 0.],
                  [0., 0., 1.]], dtype=float32)
```

```
In [133]: y_train = to_categorical(y_train)
      y_test = to_categorical(y_test)
In [134]: y_train.shape
Out[134]: (60000, 10)
In [135]: y_test.shape
Out[135]: (10000, 10)
      Define the network architecture
In [136]: # Object of neural network
      model = Sequential()
      # Input Laver
      # Hidden Layer-1
      model.add(Dense(256, activation='relu'))
      # Output layer
      model.add(Dense(10, activation='softmax'))
In [137]: model.summary()
      Model: "sequential_1"
      Layer (type)
                        Output Shape
                                         Param #
       dense_3 (Dense)
                        (None, 784)
                                         615440
       dense_4 (Dense)
                        (None, 256)
                                         200960
       dense_5 (Dense)
                        (None, 10)
                                         2570
      _____
      Total params: 818,970
      Trainable params: 818,970
      Non-trainable params: 0
      Compile the model
In [138]: model.compile(loss='categorical_crossentropy',
               optimizer='adam',
               metrics = ['accuracy'])
      Train the model
In [139]: history = model.fit(x_train, y_train, epochs=10,
            batch_size=10)
      Epoch 1/10
      6000/6000 [=
                  Epoch 2/10
      Epoch 3/10
      6000/6000 [
                     Epoch 4/10
      6000/6000 [=
               Epoch 5/10
      6000/6000 [============= ] - 19s 3ms/step - loss: 0.0457 - accuracy: 0.9859
      Epoch 6/10
      6000/6000 [=
                  Epoch 7/10
      Epoch 8/10
      6000/6000 [
```

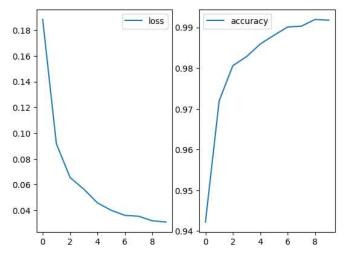
6000/6000 [==================] - 18s 3ms/step - loss: 0.0319 - accuracy: 0.9919

6000/6000 [============] - 18s 3ms/step - loss: 0.0309 - accuracy: 0.9918

Epoch 9/10

Epoch 10/10

```
In [143]: history.history
Out[143]: {'loss': [0.1883452981710434,
           0.09167750179767609,
           0.06548676639795303,
           0.056462012231349945,
           0.045726992189884186,
           0.04003522917628288,
           0.036083586513996124,
           0.03541017696261406,
           0.03188195452094078,
           0.03089030273258686],
           'accuracy': [0.942216694355011,
           0.9719499945640564,
           0.9805999994277954,
           0.9828166961669922,
           0.9858999848365784,
           0.9880499839782715,
           0.9900833368301392,
           0.9902833104133606,
           0.9919333457946777,
           0.9917500019073486]}
In [144]: plt.subplot(1,2,1)
         plt.plot(history.history['loss'], label='loss')
          plt.legend()
          plt.subplot(1,2,2)
         plt.legend()
Out[144]: <matplotlib.legend.Legend at 0x205a74a7820>
```



Out[142]: [0.1374022215604782, 0.9785000085830688]