```
1
    from google.colab import drive
 2
    drive.mount('/content/drive/')
   Mounted at /content/drive/
 1
    # import libraries
    import numpy as np
 2
 3
    import pandas as pd
    import matplotlib.pyplot as plt
 4
    from keras.models import Sequential
 5
    from keras.layers import Dense, Dropout
    from keras.utils import to_categorical
    # load the datasets for TRINING
 1
    datasets_train = pd.read_csv('/content/drive/My Drive/Colab Notebook
 2
 3
 4
    x train = datasets train.loc[:,'1x1':]
    y_train = datasets_train['label']
 5
 6
 7
    x train = np.reshape(x train, (-1, 28*28))
 8
    x train /=255
    print(x_train.head(10))
 9
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   [10 rows x 784 columns]
    # load the datasets fot TESTING
 1
    datasets_test = pd.read_csv('/content/drive/My Drive/Colab Notebooks
 2
 3
    x test = datasets test.loc[:,'1x1':]
 4
    y_test = datasets_test['label']
 5
 6
```

```
x_{\text{test}} = \text{np.resnape}(x_{\text{test}}, (-1, /84))
/
8
   x_test /=255
9
   print(x test.head(10))
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   [10 rows x 784 columns]
   # defining the keras model
1
2
   model = Sequential()
3
   # input layer
4
   model.add(Dense(1024,input_shape=(784,),activation='relu'))
   model.add(Dense(128, activation='tanh'))
5
6
   model.add(Dropout(rate=0.05))
7
   model.add(Dense(64, activation='relu'))
8
   model.add(Dense(10,activation='softmax'))
9
   # compiling the model
1
2
   model.compile(optimizer='adam',loss='categorical_crossentropy',metri
   # fitting the model
1
   history = model.fit(x=x_train,y=to_categorical(y_train),batch_size=1
  Epoch 1/5
  600/600 [============= ] - 8s 14ms/step - loss: 0.2059 - accuracy: 0.939
  Epoch 2/5
  600/600 [============ ] - 8s 14ms/step - loss: 0.0782 - accuracy: 0.97!
  Epoch 3/5
  600/600 [============ ] - 8s 14ms/step - loss: 0.0511 - accuracy: 0.98
  Epoch 4/5
  600/600 [============= ] - 8s 14ms/step - loss: 0.0396 - accuracy: 0.987
  Epoch 5/5
  600/600 [============= ] - 8s 14ms/step - loss: 0.0279 - accuracy: 0.996
```

- 1 # Evaluate the loss and accuracy of the model
- 2 loss,accuracy = model.evaluate(x=x_train,y=to_categorical(y_train),b

```
3 print(f'\nloss: {loss} \naccuracy: {accuracy} ')
```

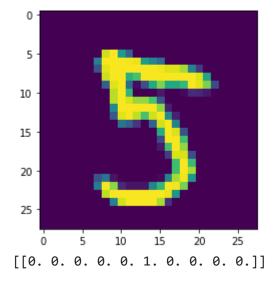
- 4 model.save('/content/drive/My Drive/Colab Notebooks/handwritten_mode
- 5 print ('model saved successfully!')

loss: 0.015853187069296837 accuracy: 0.9948833584785461

INFO:tensorflow:Assets written to: /content/drive/My Drive/Colab Notebooks/handwritten_n

model saved successfully!

```
# Make predictions
v=x_test.loc[2545,:]
a=np.reshape(np.asarray(v),(28,28))
plt.imshow(a)
plt.show()
predictions = model.predict(x=np.asarray(v).reshape((-1,784)))
print(np.round(predictions))
```



```
1
    # plot accuracy
 2
    plt.xlim(0,1000)
    plt.plot(history.history['accuracy'],'r-')
 3
    plt.title('accuracy with epochs')
 4
 5
    plt.xlabel('epochs')
    plt.ylabel('accuracy')
 6
    plt.legend(['train','test'],loc='upper right')
 7
 8
    plt.show()
 9
    # plot loss
10
    plt.plot(history.history['loss'],'r-')
11
    nlt_title('loss with enochs')
```

```
12 mlt vlabal/lamabal)
```

- 13 plt.xlabel('epochs')
- 14 plt.ylabel('loss')
- 15 plt.legend(['train','test'],loc='upper right')
- 16 plt.show()

