```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
                                                                           In [2]:
data=pd.read excel("/content/Crude Oil Prices Daily.xlsx")
                                                                           In [3]:
data.isnull().any()
                                                                         Out[3]:
Date
                 False
Closing Value
                 True
dtype: bool
                                                                           In [4]:
data.isnull().sum()
                                                                          Out[4]:
Date
                  0
Closing Value
dtype: int64
                                                                           In [5]:
data.dropna(axis=0,inplace=True)
                                                                           In [6]:
data.isnull().sum()
                                                                         Out[6]:
Date
                  0
Closing Value
dtype: int64
                                                                           In [7]:
data oil=data.reset index()['Closing Value']
data_oil
                                                                         Out[7]:
0
        25.56
        26.00
1
       26.53
       25.85
3
       25.87
        . . .
8211
       73.89
8212
       74.19
8213
       73.05
8214
       73.78
8215
        73.93
Name: Closing Value, Length: 8216, dtype: float64
                                                                           In [8]:
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler(feature range=(0,1))
data oil=scaler.fit transform(np.array(data oil).reshape(-1,1))
                                                                           In [9]:
data_oil
                                                                         Out[9]:
array([[0.11335703],
       [0.11661484],
       [0.12053902],
```

```
. . . ,
        [0.46497853],
        [0.47038353],
        [0.47149415]])
                                                                             In [10]:
plt.plot(data oil)
                                                                            Out[10]:
[]
1.0
0.8
0.6
0.4
0.2
0.0
           2000
                   4000
                           6000
training size=int(len(data oil)*0.65)
test_size=len(data_oil)-training_size
train_data,test_data=data_oil[0:training_size,:],data_oil[training_size:len
(data_oil),:1]
                                                                             In [12]:
training size, test size
                                                                            Out[12]:
(5340, 2876)
                                                                             In [13]:
train_data.shape
                                                                            Out[13]:
(5340, 1)
                                                                             In [14]:
def create_dataset(dataset, time_step=1):
  dataX, dataY=[],[]
  for i in range(len(dataset)-time step-1):
    a=dataset[i:(i+time step),0]
    dataX.append(a)
    dataY.append(dataset[i+time step,0])
  return np.array(dataX), np.array(dataY)
                                                                             In [15]:
time step=10
x_train, y_train=create_dataset(train_data, time_step)
x_test, y_test=create_dataset(test_data, time_step)
                                                                             In [16]:
print(x_train.shape),print(y_train.shape)
(5329, 10)
(5329,)
                                                                            Out[16]:
(None, None)
```

```
In [17]:
print(x test.shape),print(y test.shape)
(2865, 10)
(2865,)
                                                                    Out[17]:
(None, None)
                                                                     In [18]:
x train
                                                                    Out[18]:
array([[0.11335703, 0.11661484, 0.12053902, ..., 0.10980305, 0.1089886,
        0.11054346],
       [0.11661484, 0.12053902, 0.11550422, ..., 0.1089886 , 0.11054346,
       0.10165852],
       [0.12053902, 0.11550422, 0.1156523, ..., 0.11054346, 0.10165852,
       0.09906708],
       [0.36731823, 0.35176958, 0.36080261, ..., 0.36391234, 0.37042796,
       0.370427961,
       [0.35176958, 0.36080261, 0.35354657, ..., 0.37042796, 0.37042796,
       0.37879461],
       [0.36080261, 0.35354657, 0.35295424, ..., 0.37042796, 0.37879461,
        0.3791648211)
                                                                     In [19]:
x train=x train.reshape(x train.shape[0],x train.shape[1],1)
x test=x test.reshape(x test.shape[0],x test.shape[1],1)
                                                                     In [20]:
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import LSTM
                                                                     In [22]:
model=Sequential()
                                                                     In [23]:
model.add(LSTM(50, return sequences=True, input shape=(10,1)))
model.add(LSTM(50, return sequences=True))
model.add(LSTM(50))
model.add(Dense(1))
                                                                     In [25]:
model.summary()
Model: "sequential 1"
Layer (type)
                           Output Shape
                                                     Param #
______
                            (None, 10, 50)
lstm (LSTM)
                                                      10400
lstm 1 (LSTM)
                            (None, 10, 50)
                                                      20200
lstm 2 (LSTM)
                            (None, 50)
                                                      20200
dense (Dense)
                            (None, 1)
                                                      51
```

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Total params: 50,851

```
Trainable params: 50,851
Non-trainable params: 0
```

```
In [26]:
model.compile(loss='mean squared error',optimizer='adam')
                                                        In [27]:
model.fit(x_train, y_train, validation_data=(x_test, y_test), epochs=3, batch si
ze=64, verbose=1)
Epoch 1/3
84/84 [============== ] - 11s 48ms/step - loss: 0.0023 - val
loss: 0.0010
Epoch 2/3
val loss: 8.0346e-04
Epoch 3/3
val_loss: 9.4227e-04
                                                        Out[27]:
                                                        In [28]:
train predict=scaler.inverse transform(train data)
test_predict=scaler.inverse_transform(test data)
### Calculate RMSE performance metrics
import math
from sklearn.metrics import mean squared error
math.sqrt(mean_squared_error(train_data,train_predict))
                                                        Out[28]:
```

29.347830443269938