## SAVE THE MODEL

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In [3]:
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
 In [4]: data=pd.read_excel("/content/Crude Oil Prices Daily.xlsx")
 In [5]: data.isnull().any()
Out[5]: Date False
Closing Value dtype: bool
In [6]: data.isnull().sum()
Out[6]: Date 0
Closing Value 7
dtype: int64
In [7]: data.dropna(axis=0,inplace=True)
In [8]: data.isnull().sum()
Out[8]: Date 0
Closing Value 0
dtype: int64
Out[9]: 0 25.56
1 26.00
Out[9]: 0 25.56
1 26.00
2 26.53
3 25.85
4 25.87
           21.07
8211 73.89
8212 74.19
8213 73.05
8214 73.78
8215 73.93
Name: Closing Value, Length: 8216, dtype: float64
In [10]:
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler(feature_range=(0,1))
data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))
In [11]: data_oil
...,
[0.46497853],
[0.47038353],
[0.47149415]])
In [12]: plt.plot(data_oil)
Out[12]: []
           0.8
```

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Out[12]: []
              1.0
              0.8
               0.6
               0.4
               0.2
               0.0
              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 [14]: training_size,test_size
 Out[14]: (5340, 2876)
 In [15]: train_data.shape
 Out[15]: (5340, 1)
In [17]:
            time_step=10
            x_train_x_train=create_dataset(train_data,time_step)
            x_test,y_test=create_dataset(test_data,time_step)
 In [18]: print(x_train.shape),print(y_train.shape)
             (5329, 10)
(5329,)
Out[18]: (None, None)
 In [19]: print(x_test.shape),print(y_test.shape)
             (2865, 10)
(2865,)
Out[19]: (None, None)
In [20]: x_train
Out[20]: 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],
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Out[20]: array([[0.11335703, 0.11661484, 0.12053902, ..., 0.10980305, 0.1089886 ,
                               (0.11054346),

[0.11061484, 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, 0.37042796], [0.55176958, 0.36080261, 0.35354657, ..., 0.37042796, 0.37042796, 0.37042796, [0.35516957, 0.35295424, ..., 0.37042796, 0.37879461, 0.37916482]])
In [21]:  \begin{array}{ll} x \text{ train=x\_train.reshape}(x \text{ train.shape}[\theta], x \text{ train.shape}[1], 1) \\ x \text{ _test=x\_test.reshape}(x \text{ _test.shape}[\theta], x \text{ _test.shape}[1], 1) \end{array} 
                     from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense from tensorflow.keras.layers import LSTM
In [23]: model=Sequential()
In [24]:
    model.add(LSTM(50,return_sequences=True,input_shape=(10,1)))
    model.add(LSTM(50,return_sequences=True))
    model.add(LSTM(50))
In [25]: model.add(Dense(1))
In [26]: model.summary()
                   Model: "sequential"
                                                                      Output Shape Paraim +
                  Layer (type)
                   lstm (LSTM)
                   lstm_1 (LSTM)
                                                                    (None, 10, 50)
                                                                                                                        20200
                                                                   (None, 50)
                   lstm_2 (LSTM)
                                                                                                                        20200
                   dense (Dense)
                                                                       (None, 1)
                  Total params: 50,851
Trainable params: 50,851
Non-trainable params: 0
In [27]: model.compile(loss='mean_squared_error',optimizer='adam')
\label{eq:model.fit} \mbox{In [28]:} \quad \mbox{model.fit}(x\_\mbox{train},y\_\mbox{train},\mbox{validation}\_\mbox{data}=(x\_\mbox{test},y\_\mbox{test}),\mbox{epochs}=3,\mbox{batch}\_\mbox{size}=64,\mbox{verbose}=1)
                  Epoch 1/3
                 Epoch 1/3
84/98 [===========] - 10s 49ms/step - loss: 0.0018 - val_loss: 0.0010
Epoch 2/3
84/98 [=============] - 3s 30ms/step - loss: 1.2823e-04 - val_loss: 7.6827e-04
Epoch 3/3
84/98 [==============] - 3s 32ms/step - loss: 1.2320e-04 - val_loss: 7.7520e-04
                   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[29]: 29.347830443269938