## MODEL BUILDING- MODEL EVALUATION

Team ID	PNT2022TMID09674
Project Name	Crude Oil Price Prediction

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In [1]:
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
Closing Value
dtype: bool
In [4]: data.isnull().sum()
Out[4]: Date
           Closing Value
dtype: int64
In [5]: data.dropna(axis=0,inplace=True)
In [6]: data.isnull().sum()
Out[6]: Date
Closing Value
           dtype: int64
In [7]:
    data_oil=data.reset_index()['Closing Value']
    data_oil
                    25.56
26.00
26.53
25.85
Out[7]: 0
1
            4
                     25.87
                   73.89
74.19
73.05
73.78
73.93
            8211
            8212
8213
8214
8215
            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
 ...,
[0.46497853],
                     [0.47038353],
[0.47149415]])
In [10]: plt.plot(data_oil)
```

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Out[10]: []
               1.0
                0.6
                0.4
                0.2
In [11]:
    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)
               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 [14]:
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
```

```
\texttt{Out[18]: array([[0.11335703,\ 0.11661484,\ 0.12053902,\ \dots,\ 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],
                 [ \tt 0.36731823, \ \tt 0.35176958, \ \tt 0.36080261, \ \ldots, \ \tt 0.36391234, \ \tt 0.37042796, 
               0.37042796],
[0.35176958, 0.36080261, 0.35354657, ..., 0.37042796, 0.37042796,
0.37879461],
[0.36080261, 0.35354657, 0.35295424, ..., 0.37042796, 0.37879461,
                0.37916482]])
         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))
In [24]: model.add(Dense(1))
In [25]: model.summary()
         Model: "sequential_1"
   Layer (type)
                                                   Output Shape
                                                                                                Param #
  ______
   1stm (LSTM)
                                                  (None, 10, 50)
                                                                                               10400
   lstm_1 (LSTM)
                                                  (None, 10, 50)
                                                                                               20200
   1stm 2 (LSTM)
                                                   (None, 50)
                                                                                                20200
   dense (Dense)
                                                   (None, 1)
                                                                                                51
 ______
 Total params: 50,851
 Trainable params: 50,851
 Non-trainable params: 0
          model.compile(loss='mean_squared_error',optimizer='adam')
 \label{eq:model.fit} \begin{tabular}{ll} In [27]: & model.fit(x\_train,y\_train,validation\_data=(x\_test,y\_test),epochs=3,batch\_size=64,verbose=1) \\ \end{tabular}
         Epoch 1/3
84/84 [===
Epoch 2/3
                         -----] - 11s 48ms/step - loss: 0.0023 - val_loss: 0.0010
         84/84 [------] - 2s 28ms/step - loss: 1.2599e-04 - val_loss: 8.0346e-04
Epoch 3/3
```

========] - 2s 30ms/step - loss: 1.2479e-04 - val\_loss: 9.4227e-04

import math

train\_predict=scaler.inverse\_transform(train\_data)
test\_predict=scaler.inverse\_transform(test\_data)

from sklearn.metrics import mean\_squared\_error
math.sqrt(mean\_squared\_error(train\_data,train\_predict))

### Calculate RMSE performance metrics

In [28]: