MODEL BUILDING-CONFIGURING THE LEARNING PROCESS

Team ID	PNT2022TMID52922
Project Name	Crude Oil Price Prediction

```
import pandas as pd
     import numpy as np
import matplotlib.pyplot as plt
[ ] data = pd.read_excel("/content/drive/MyDrive/Crude Oil Prices Daily.xlsx")
     data.head()
             Date Closing Value
      0 1986-01-02 25.56
      1 1986-01-03
                             26.00
     2 1986-01-06 26.53
      3 1986-01-07
      4 1986-01-08 25.87
[ ] data.isnull().any()
     Date False
Closing Value True
     dtype: bool
[ ] data.dropna(axis=0,inplace=True)
[ ] data.isnull().sum()
     Date 8
Closing Value 8
dtype: int64
data_oil-data.reset_index()['Closing Value']
data_oil
            25.56
26.00
26.53
25.85
25.87
0
     * 23.87

8211 73.89

8212 74.19

8213 73.65

8214 73.78

8215 73.93

Name: Closing Value, Length: 8216, dtype: float64
[ ] from sklearn.preprocessing import MinMaxScaler
     scaler=MinMaxScaler(feature_range=(0,1))
     data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))
[] data_oil
     array([[0.11335783],
            [0.11661484],
[0.12053902],
            [0.46497853],
[0.47838353],
[0.47149415]])
[ ] plt.plot(data_oil)
```

```
[cmatplotlib.lines.Line2D at 0x7f05a724c690x]
        0.6
        0.2
[ ] 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]
[ ] training_size,test_size
      (5348, 2876)
[ ] train_data.shape
      (5348, 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)
[] time_step=10
       x_train,y_train-create_dataset(train_data,time_step)
x_test,y_test-create_dataset(test_data,time_step)
 print(x_train.shape),print(y_train.shape)
 (5329, 10)
(5329,)
[ ] print(x_test.shape),print(y_test.shape)
[] x_train
      array([[0.11335783, 0.11661484, 0.12853902, ..., 0.10988385, 0.1089886, 0.11054346],
[0.11651484, 0.12053902, 0.11550422, ..., 0.1089886, 0.11054346, 0.10165852],
[0.12053902, 0.11550422, 0.1156523, ..., 0.11054346, 0.10165852, 0.09906708],
                ...,
[8.36731823, 8.35176958, 8.36888261, ..., 8.36391234, 8.37842796,
               [0.36/31823, 0.35176998, 0.36888251, ..., 0.35391234, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37642795, 0.37879461, 0.37916482]])
[ ] x_train=x_train.reshape(x_train.shape[0],x_train.shape[1],1)
       \texttt{x\_test-x\_test.reshape}(\texttt{x\_test.shape}[\theta],\texttt{x\_test.shape}[1],\texttt{1})
[ ] from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import LSTM
 [ ] model=Sequential()
  [ ] 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))
  model.summary()
   Model: "sequential"
                                                     Output Shape
          Layer (type)
          1stm (LSTM)
                                                     (None, 10, 58)
                                                                                             18488
          lstm_1 (LSTM)
                                                     (None, 10, 50)
                                                                                             28288
          1stm_2 (LSTM)
                                                    (None, 58)
                                                                                             28288
          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')