

```
In [2]: import requests
import io
url = r"https://github.com/IBM-EPBL/IBM-Project-51555-1660980405/blob/main/Data%20Collection/DataSet1986-2018.xlsx?raw=true"
download = requests.get(url).content
data = pd.read_excel(url,index_col=0,parse_dates=[0])
print(data.head())
```

	Closing Value
Date	
1986-01-02	25.56
1986-01-03	26.00
1986-01-06	26.53
1986-01-07	25.85
1986-01-08	25.87

In []:

```
In [3]: data.isnull().any()
data.isnull().sum()
data.dropna(axis = 0, inplace = True)
data.isnull().sum()
data_oil = data.reset_index()['Closing Value']
data_oil
```

```
Out[3]: 0      25.56
1      26.00
2      26.53
3      25.85
4      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 [4]: from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler (feature_range=(0,1))
data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))
```

Projects / Crude Oil Price Prediction / Model

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Epoch 41/50
84/84 [=====] - 3s 42ms/step - loss: 4.0582e-05 - val_loss: 1.9684e-04
Epoch 42/50
84/84 [=====] - 3s 35ms/step - loss: 4.0487e-05 - val_loss: 1.9967e-04
Epoch 43/50
84/84 [=====] - 3s 34ms/step - loss: 4.4339e-05 - val_loss: 2.0307e-04
Epoch 44/50
84/84 [=====] - 3s 30ms/step - loss: 3.7815e-05 - val_loss: 2.1742e-04
Epoch 45/50
84/84 [=====] - 3s 33ms/step - loss: 3.6488e-05 - val_loss: 2.2140e-04
Epoch 46/50
84/84 [=====] - 3s 34ms/step - loss: 3.6334e-05 - val_loss: 2.1881e-04
Epoch 47/50
84/84 [=====] - 3s 38ms/step - loss: 4.1092e-05 - val_loss: 4.8617e-04
Epoch 48/50
84/84 [=====] - 3s 31ms/step - loss: 3.6932e-05 - val_loss: 2.4415e-04
Epoch 49/50
84/84 [=====] - 3s 33ms/step - loss: 3.5177e-05 - val_loss: 1.8624e-04
Epoch 50/50
84/84 [=====] - 3s 33ms/step - loss: 3.6481e-05 - val_loss: 2.0401e-04
```

Out[13]: <keras.callbacks.History at 0x7f7341326d30>

```
In [14]: train_predict = model.predict(X_train)
test_predict = model.predict(X_test)
train_predict = scaler.inverse_transform(train_predict)
test_predict = scaler.inverse_transform(test_predict)

import math
from sklearn.metrics import mean_squared_error
math.sqrt(mean_squared_error(y_train,train_predict))
```

Out[14]: 29.71311299049386

```
In [15]: from tensorflow.keras.models import load_model
model.save("crude_oil.h5")
```

```
In [16]: look_back=10
trainpredictPlot = np.empty_like(data_oil)
trainpredictPlot[:, :] = np.nan
trainpredictPlot[look_back:len(train_predict)+look_back, :] = train_predict
```