FINAL CODE

Team ID	PNT2022TMID28643
Project Name	Crude Oil Price Prediction

Source Code

Building the model:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

data = pd.read_excel("Crude Oil Prices Daily.xlsx")
data.head()

data.isnull().any()

data.isnull().sum()

data.dropna(axis=0,inplace=True)
data.isnull().sum()

data_oil = data.reset_index()["Closing Value"]
data_oil

from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler ( feature_range = (0,1) )
data_oil = scaler.fit_transform(np.array(data_oil).reshape(-1,1))

plt.title('Crude OIl Price')
plt.plot(data_oil)

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

train_data.shape

import numpy
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, ytest = create_dataset(test_data, time_step)

print(X_train.shape), print(y_train.shape)

print(X_test.shape), print(ytest.shape)

X_train

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)

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.compile(loss='mean_squared_error', optimizer = 'adam')

model.fit(X_train, y_train, validation_data = (X_test, ytest), epochs = 10, batch_size = 64,
verbose = 1)

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))

from tensorflow.keras.models import load_model
model.save("Crude_oil.h5")

look_back = 0
trainPredictPlot = np.empty_like(data_oil)
trainPredictPlot[:, :] = np.nan
trainPredictPlot[look_back:len(train_predict) + look_back, :] = train_predict

testPredictPlot = np.empty_like(data_oil)
testPredictPlot[:,:] = np.nan
testPredictPlot[len(train_predict)+(look_back*2)+1: len(data_oil)-1, :] = test_predict

plt.plot(scaler.inverse_transform(data_oil))
```

```
plt.plot(trainPredictPlot)
plt.plot(testPredictPlot)
plt.title("Testing The Model")
plt.show()

len(test_data)

x_input = test_data[2866:].reshape(1,-1)
x_input.shape

temp_input = list(x_input)
temp_input = temp_input[0].tolist()
temp_input

lst_output = []
n_steps = 10
i=0
while(i<10):
    if(len(temp_input)>10):
        x_input = np.array(temp_input[1:])
        print("{} day input {}".format(i,x_input))
        x_input = x_input.reshape(1,-1)
        x_input = x_input.reshape((1,n_steps, 1))

        yhat = model.predict(x_input, verbose = 0)
        print("{} day output {}".format(i,yhat))
        temp_input.extend(yhat[0].tolist())
        temp_input = temp_input[1:]
        lst_output.extend(yhat.tolist())
        i=i+1

    else:
        x_input = x_input.reshape((1, n_steps,1))
        yhat = model.predict(x_input, verbose = 0)
```

```
        print(yhat[0])
        temp_input.extend(yhat[0].tolist())
        print(len(temp_input))
        lst_output.extend(yhat.tolist())
        i=i+1

day_new = np.arange(1,11)
day_pred = np.arange(11,21)

len(data_oil)

plt.plot(day_new,scaler.inverse_transform(data_oil[8206:]))
plt.title("Review Of Prediction")
plt.plot(day_pred,scaler.inverse_transform(lst_output))
plt.show()

df3 = data_oil.tolist()
df3.extend(lst_output)
plt.title("Past Data & Next 10 Days Output Prediction")
plt.plot(df3[8100:])

df3 = scaler.inverse_transform(df3).tolist()
plt.title("Past Data & Next 10 Days Output Prediction After Reversing The Scaled Values")
plt.plot(df3)
```

Deploying on IBM Cloud:

```
get_ipython().system('pip install ibm_watson_machine_learning')

from ibm_watson_machine_learning import APIClient
wml_credentials = {
   "url": "https://us-south.ml.cloud.ibm.com",
   "apikey": "uVEty-CB4dYcccQ_Jq9V-atVXmL1dByE_wiDm95lcyTQ"
}
```

```
client = APIClient(wml_credentials)

def guid_from_space_name(client, NewSpace):
   space = client.spaces.get_details()
   return(next(item for item in space['resources'] if item['entity']["name"] ==
NewSpace)['metadata']['id'])

space_uid = guid_from_space_name(client, 'NewSpace')
print("Space UID = " + space_uid)

client.set.default_space(space_uid)

client.software_specifications.list()

software_spec_id = client.software_specifications.get_id_by_name('tensorflow_rt22.1-
py3.9')
print(software_spec_id)

model.save('crude.h5')

get_ipython().system('tar -zcvf crude-oil.tgz Crude.h5')

software_space_uid = client.software_specifications.get_uid_by_name('tensorflow_rt22.1-
py3.9')
software_space_uid

model_details = client.repository.store_model(model='crude.tgz',meta_props={
client.repository.ModelMetaNames.NAME:"crude_oil_model",
client.repository.ModelMetaNames.TYPE:"tensorflow_2.7",
client.repository.ModelMetaNames.SOFTWARE_SPEC_UID:software_spec_id }
                                  )
model_id = client.repository.get_model_uid(model_details)
model_id
```

client.repository.download(model_id,'crude_oil_model.tar.gb')

INTEGRATE FLASK WITH SCORING END POINT

App.py

```
from flask import Flask,render_template,request,redirect
import pandas as pd
import numpy as np
from flask import Flask, render_template, Response, request
import pickle
from sklearn.preprocessing import LabelEncoder
import requests

# NOTE: you must manually set API_KEY below using information retrieved from your
IBM Cloud account.
API_KEY = "uVEty-CB4dYcccQ_Jq9V-atVXmL1dByE_wiDm95lcyTQ"
token_response = requests.post('https://iam.cloud.ibm.com/identity/token',
data={"apikey":API_KEY, "grant_type": 'urn:ibm:params:oauth:grant-type:apikey'})
mltoken = token_response.json()["access_token"]
header = {'Content-Type': 'application/json', 'Authorization': 'Bearer ' + mltoken}


app = Flask(_name_)

@app.route('/',methods=["GET"])
def index():
    return render_template('index.html')


@app.route('/predict',methods=["POST","GET"])
def predict():
    if request.method == "POST":
        string = request.form['val']
```

```
string = string.split(',')
temp_input = [eval(i) for i in string]

x_input = np.zeros(shape=(1, 10))
x_input.shape

lst_output = []
n_steps = 10
i=0
while(i<10):
    if(len(temp_input)>10):
        x_input = np.array(temp_input[1:])
        x_input = x_input.reshape(1,-1)
        x_input = x_input.reshape((1,n_steps, 1))
        yhat = model.predict(x_input, verbose = 0)
        temp_input.extend(yhat[0].tolist())
        temp_input = temp_input[1:]
        lst_output.extend(yhat.tolist())
        i=i+1

    else:
        x_input = x_input.reshape((1, n_steps,1))
        yhat = model.predict(x_input, verbose = 0)
        temp_input.extend(yhat[0].tolist())
        lst_output.extend(yhat.tolist())
        i=i+1


    # NOTE: manually define and pass the array(s) of values to be scored in the next line
    payload_scoring = {"input_data": [{   "values": [[x_input]]    }]}

    response_scoring = requests.post('https://us-
south.ml.cloud.ibm.com/ml/v4/deployments/7f67cbed-6222-413b-9901-
```

```
b2a72807ac82/predictions?version=2022-10-30', json=payload_scoring,
headers={'Authorization': 'Bearer ' + mltoken})
        predictions = response_scoring.json()
        print(response_scoring.json())

        val = lst_output[9]
        return render_template('web.html' , prediction = val)



  if request.method=="GET":
     return render_template('web.html')

if_name_=="_main__":
  model = load_model('C:/Users/rkara/IBM/Sprint - 4/Crude_oil.tar.gz')
  app.run(debug=True)
```

INDEX.HTML

```
<!DOCTYPE html>
<head>
  <title>Crude Oil Price Prediction </title>
  <link rel="stylesheet" href="{{ url_for('static', filename='css/index.css') }}">
</head>
<body>
  <h1> Crude Oil Price Prediction</h1>
  <p> Demand for oil is inelastic, therefore the rise in price is good news
   for producers because they will see an increase in their revenue. Oil
   importers, however, will experience increased costs of purchasing oil.
   Because oil is the largest traded commodity, the effects are quite
   significant. A rising oil price can even shift economic/political
   power from oil importers to oil exporters. The crude oil price movements
   are subject to diverse influencing factors.
  </p><br><br>
  <a href="{{url_for('predict')}}">
```

```
    Predict Future Price</a>
</body>
```

WEB.HTML

```
<!DOCTYPE html>
<head>
    <title>Crude Oil Price Prediction </title>
    <link rel="stylesheet" href="{{ url_for('static', filename='css/web.css') }}">
</head>
<body>
    <h1>
    Crude Oil Price Prediction </h1>
    <form action="/predict" method="POST" enctype = "multipart/form-data">
        <input type="text" name="val" placeholder="Enter the crude oil price for first 10 days"
>
         <br> <br> <br>
        <input type="submit"/>
    </form><br> <br>
    <div>
        {{prediction}}
    </div>

</body>
```