# **CRUDE OIL PREDICTION**

### **PROJECT REPORT**

Team Id: PNT2022TMID00817

Submitted by

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# **Panimalar Engineering College**



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#### 1. Introduction

### 1.1 Project Overview

Due to the inelasticity of oil demand, the rise in price will increase revenue for producers. However, oil importers will have to pay more for their oil purchases. Due to the fact that oil is the most traded commodity, the effects are quite significant. Rising oil prices can even shift economic/political power from oil importers to oil exporters. Diverse factors influence crude oil price movements.

The purpose of this Guided Project is to apply Neural Networks to predict the Crude Oil Price. By making this decision, we are able to buy crude oil at the right time. This kind of prediction is best made using time series analysis since we are predicting crude oil prices based on their past history. As a result, we will implement RNNs (Recurrent Neural Networks) with LSTMs (Long Short Term Memory).

Price forecasting can help minimise risks associated with volatility in oil prices by reducing the impact of price fluctuations on global economies. Investors, policymakers, and governments all rely on price forecasts for various purposes

### 1.2 Purpose

The cost of crude oil affects the global habitat, our economy, oil exploration, exploitation, and other activities directly, making it one of the most important resources in today's world. Many large and small industries, individuals, and governments rely on the prediction of oil prices, which has become a necessity.

Due to crude oil's evaporative nature, predicting its price becomes extremely difficult and inaccurate. Crude oil prices are affected by a variety of factors. By using artificial neural networks (ANNs), we propose a contemporary and innovative method for predicting crude oil prices.

This approach is able to continuously capture the unstable pattern of crude oil prices that has been incorporated by finding the optimal lag and number of the delay effect that controls crude oil prices.

After varying the lag over time for close and optimum results, we evaluated the root mean square error to validate our results. The results obtained using the proposed model were significantly better than those obtained using the previous model.

### 2. Literature Survey

### 2.1 Existing Problem

Crude oil, one of the world's most important commodities, accounts for one-third of global energy consumption. From transportation fuels to plastics, it is a starting material for most products we use every day. Price forecasting can assist in minimizing risks associated with oil price volatility by reducing the impact of fluctuations on global economies.

Various stakeholders, including governments, public and private enterprises, policymakers, and investors, rely heavily on price forecasts. From the equilibrium between demand and supply, crude oil's price should be predictable based on economic theory, where demand forecasts are usually derived from GDP, exchange rates, and domestic prices, while supply forecasts are derived from past production and reserve data.

Usually, oil demand can be predicted easily, but supply is heavily affected by political activity, such as cartelization by OPEC to regulate prices, technological advancements that allow more oil to be extracted, and wars and other conflicts that can affect supply unpredictably.

Economic models that incorporate supply and demand parameters and their determinants are referred to as structural models. Despite structural models being the most logical way to model industrial product prices, crude oil is influenced by many other factors. Among these factors is the fact that crude oil's price is determined by the futures market, which allows the purchase of a predefined amount of oil at a specified price in the future.

#### 2.2 References

- 1. Sehgal, N.; Pandey, K.K. Artificial intelligence methods for oil price forecasting: A review and evaluation. Energy Syst.2015,6,479–506.
- 2. Shuang Gao, Yalin Lei. "A new approach for crude oil price prediction based on stream learning", Geoscience Frontiers, 2017
- 3. Bashiri Behmiri, Niaz and Pires Manso, José Ramos, Crude Oil Price Forecasting Techniques: A Comprehensive Review of Literature
- Kulkarni, S., Haidar, I., 2009. Forecasting model for crude oil price using artificial neural networks and commodity future prices. International Journal of Computer Science and Information Security 2 (1)
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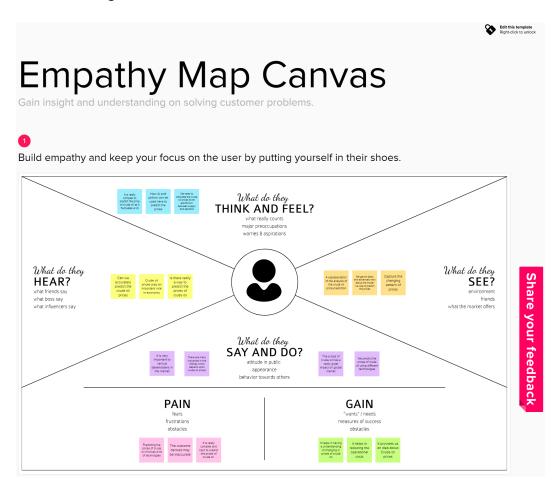
#### 2.3 Problem Statement Definition

Oil demand 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

### 3. Ideation & Proposed Solution

### 3.1 Empathy Map Canvas

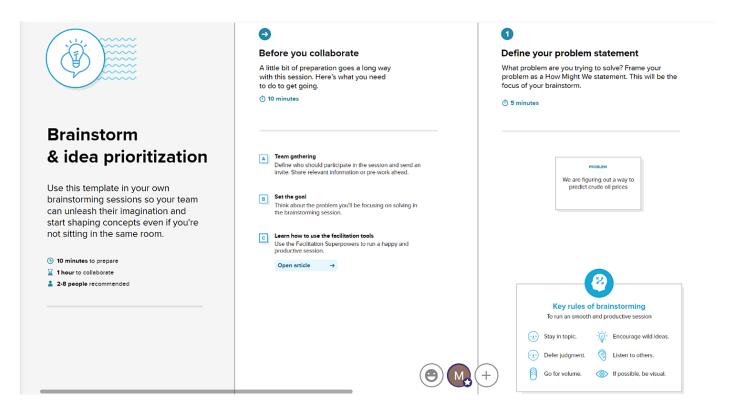
Empathy maps are simple easy-to-digest visuals that convey knowledge about a user's attitude and behavior. Team members can use it to better understand their users with the help of this tool. Understanding the true problem and the person experiencing it is essential for creating an effective solution. Through the process of creating the map, participants gain a better understanding of how the user sees the world, his or her goals, and the challenges he or she faces.



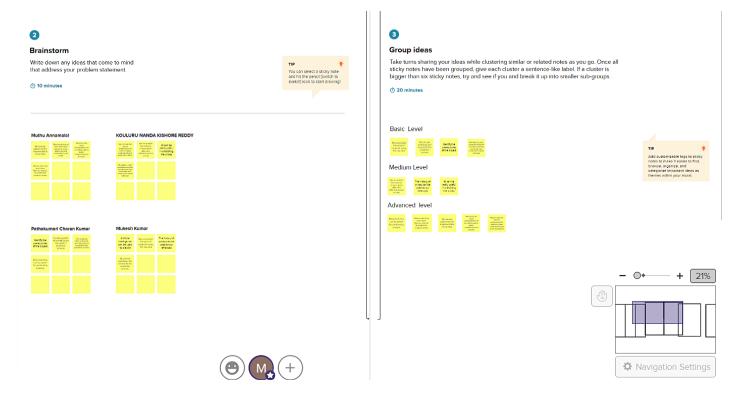
### 3.2 Ideation & Brainstorming

Brainstorming fosters a creative thinking process that leads to problem solving by providing a free and open environment. We prioritise volume over value, welcome out-of-the-box ideas, and encourage all participants to collaborate in order to develop a rich amount of creative solutions.

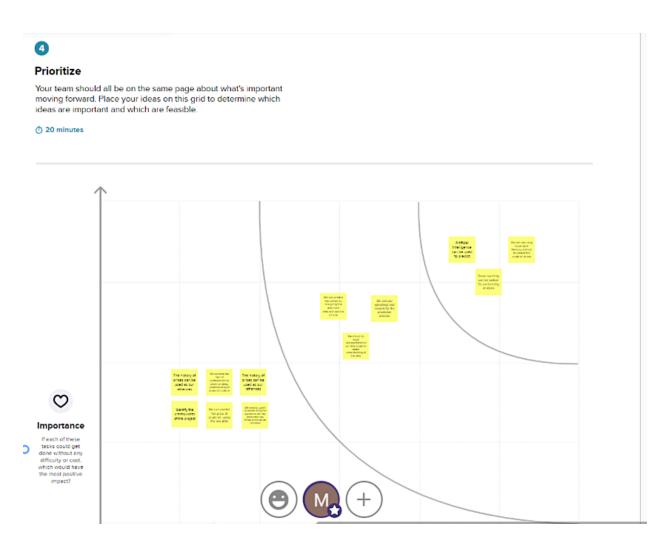
Step 1: Team Gathering, Collaboration and Select the Problem Statement



## Step 2: Brainstorm, Idea Listing and Grouping



## **Step 3: Idea Prioritization**



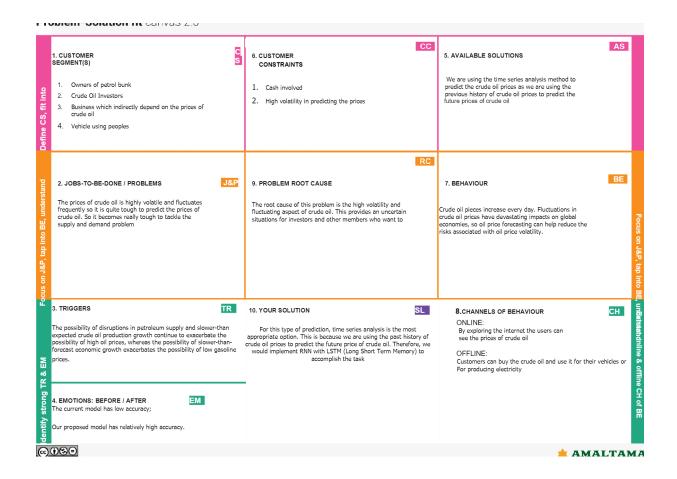
# **3.3 Proposed Solution**

| S.NO | Parameter                                      | Description  |
|------|--|--|
| 1.   | Problem Statement<br>(Problem to be<br>solved) | Oil prices play a majorrole in determining the global economies but determining it is really tough. We are trying to solve this  |
| 2.   | Idea/Solution<br>description                   | The crude oil price can be easilypredicted by usingthe counterbalance between supply, and demand using past data. We are predicting it using Python, RNN, and Deep learning. We have triedto incorporate timeseries analysis methodto predict the prices of crude oil  |
| 3.   | Novelty / Uniqueness                           | The main objective of our project is to apply Neural Networks to predict crude oil prices. By making this decision, we can buy crude oil at the righttime. In orderto make this kind of prediction, time series analysis is the best option since we are using the past history of crude oil prices to make predictions about the future. To accomplish the task, we would implement RNN (Recurrent Neural Network) with LSTM (LongShort Term Memory). |
| 4.   | Social Impact/<br>Customer Satisfaction        | Oil price changes have important consequences for the global economy, so we try to use our model to predict it in order to help the economy and businesses aroundthe world.  |
| 5.   | Business Model<br>(Revenue Model)              | Our revenuemodel is focused on  1. Pay per month model  2. Pay per year model  |

| 6. | Scalability of the | The time series analysis method is used to predict   |
|----|--------------------|--|
|    | Solution           | crude oil prices on the basis of previous historical |
|    |                    | data. We believe that we can provide better and      |
|    |                    | more accurate predictions of crude oil prices, so    |
|    |                    | we tend to satisfy the customer, which in turn       |
|    |                    | makes us more scalable.                              |

#### 3.4 Problem Solution Fit

A Problem-Solution Fit is simply the finding of a problem in your customer's life and the realization of a solution for it that actually solves the problem. Entrepreneurs, marketers, and corporate innovators can identify behavioral patterns and recognize what works and why



# 4. Requirement Analysis

## **4.1 Functional Requirements**

Following are the functional requirements of the proposed solution.

| FR No. | Functional         | Sub Requirement (Story/ Sub-Task) |
|--------|--------------------|-----------------------------------|
|        | Requirement (Epic) |                                   |
| FR-1   | User Registration  | Registration through              |
|        |                    | Mobile number                     |
|        |                    | Registration through              |
|        |                    | Gmail                             |
| FR-2   | User Confirmation  | Confirmation via Email            |
|        |                    | Confirmation via OTP              |
| FR-3   | login              | User can login through registered |
|        |                    | email ID/Mobilenumber             |

# **4.2 Non-Functional Requirements**

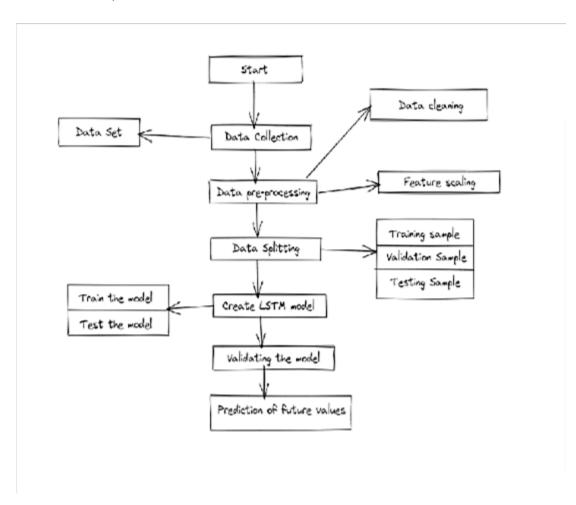
Following are the non-functional requirements of the proposed solution.

| FR   | Non-Functional | Description   |  |  |  |  |
|------|----------------|---|--|--|--|--|
| No.  | Requirement    |   |  |  |  |  |
| NFR- | Usability      | UI is user-friendly .we represent                                     |  |  |  |  |
| 1    |                | the data in charts whichuses clear                                    |  |  |  |  |
|      |                | understanding of price activity  We follow certain security protocols |  |  |  |  |
| NFR- | Security       | We follow certain security protocols                                  |  |  |  |  |
| 2    |                | likeusing user  |  |  |  |  |
|      |                | credentials , OTP verification  |  |  |  |  |
| NFR- | Reliability    | The Data which represented in   |  |  |  |  |
| 3    |                | web app is so accurate and  |  |  |  |  |
|      |                | predicting the right data   |  |  |  |  |
| NFR- | Performance    | The performance in this projectis                                     |  |  |  |  |
| 4    |                | determined through "how accurately                                    |  |  |  |  |
|      |                | you can predict the price of the crude                                |  |  |  |  |
|      |                | oil "   |  |  |  |  |
| NFR- | Availability   | The web appis available to all  |  |  |  |  |
| 5    |                | devices(Android , Mac   |  |  |  |  |
|      |                | , windows etc.,)  |  |  |  |  |
| NFR- | Scalability    | According to user basethe project                                     |  |  |  |  |
| 6    |                | scalability is done   |  |  |  |  |

## 5. Project Design

#### 5.1 Data Flow Diagram

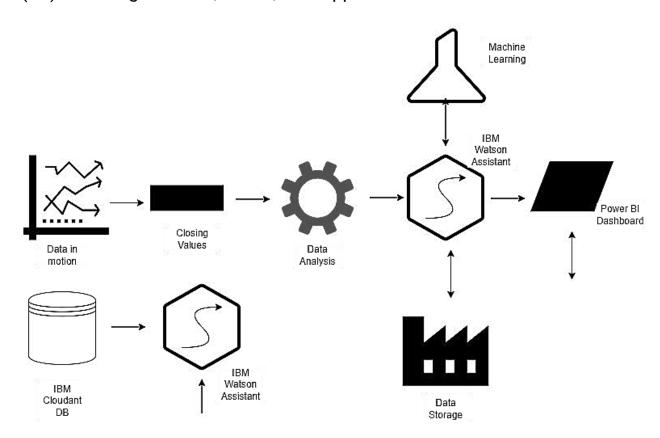
Data Flow Diagrams(DFDs) are traditional visual representations of how information flows within a system. It is possible to illustrate the right amount of the system requirement graphically with a neat and clear DFD. It explains how data enters and leaves the system, what changes the information, and where it is stored.



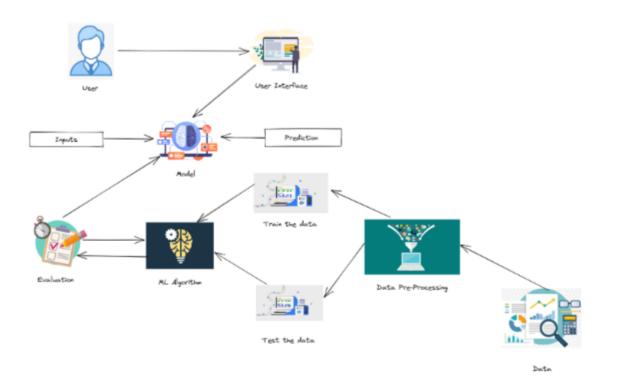
#### 5.2 Solution & Technical Architecture

An architectural description of a solution is called a Solution architecture (SA). Enterprise solution architectures (ESAs) combine guidance from various enterprise architecture viewpoints (business, information, and technical). Solution architecture aims to achieve the following overarching goals:

- (i) Streamlining day-to-day operations
- (ii) Providing a more efficient production environment
- (iii ) Lowering costs and gainingcost-effectiveness
- (iv) Providing a secure, stable, and supportable environment



**Solution Architecture** 



**Technical Architecture** 

## **5.3 User Stories**

| Sprint   | Functional<br>Requireme<br>nt (Epic) | User<br>Story<br>Num<br>ber | User Story / Task  |
|----------|--------------------------------------|-----------------------------|--|
| Sprint-1 | Data<br>Collection                   | USN-1                       | Collecting the Dataset   |
| •        | DataPre-<br>processing               | USN-2                       | Data Pre-processing  |
| Sprint-3 | Model<br>Building                    | USN-3                       | Prepare the model by importing thenecessary libraries, adding thelayers, and compiling it. |
| Sprint-3 | Model<br>Building                    | USN-4                       | The data classification model is trained using RNNs and other systems.                     |

| Sprint-4 | Application<br>Building | USN-5 | Deploy the modelin the IBMcloudand build the system |
|----------|-------------------------|-------|---|
| Sprint-4 | Training and testing    | USN-6 | Testing the model's performanceand training it      |

# 6. Project Planning & Scheduling

# **6.1 Sprint Planning & Estimation**

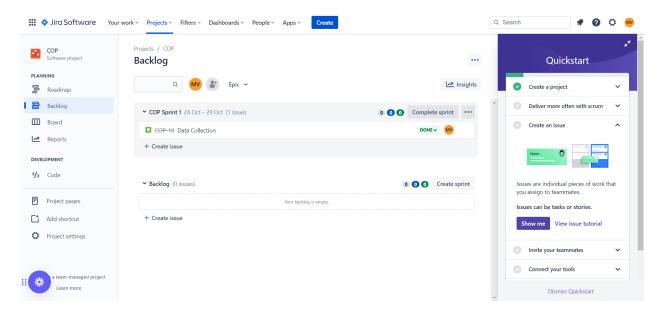
| Sprint       | Function<br>al<br>Require<br>ment<br>(Epic) | Us<br>er<br>Sto<br>ry<br>Nu<br>mb<br>er | User Story /<br>Task  | Sto<br>ry<br>Poi<br>nts | Priori<br>ty | Team Members   |
|--------------|---|---|---|-------------------------|--------------|--|
| Sprint-      | Data<br>Collection                          |   | Collecting the<br>Dataset   | 10                      | 3            | Muthu Annamalai<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar<br>Charan Kumar |
| Sprint-<br>2 | Data Pre-<br>processing                     |   | Data Pre-<br>processing   | 7                       |              | Muthu Annamalai<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar<br>Charan Kumar |
| Sprint-3     | Model<br>Building                           |   | Prepare the model by importing thenecessary libraries, adding the layers, and compiling it. | 10                      | 3            | Muthu Annamalai<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar<br>Charan Kumar |

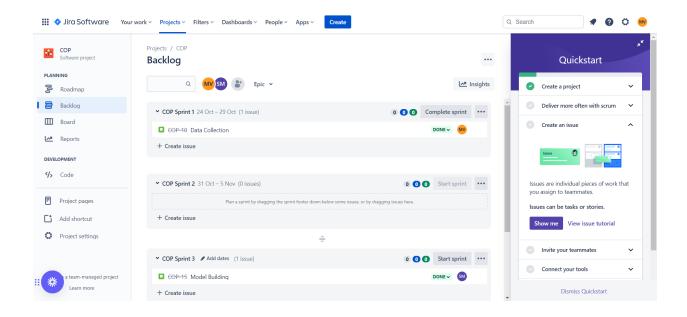
| Sprint-  | Model<br>Building           | USN-4 | The data classificati on model is trained using RNNs and other systems. | 7  | Medi<br>um | Muthu Annamalai<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar<br>Charan Kumar     |
|----------|-----------------------------|-------|---|----|------------|--|
| Sprint-4 | Applicati<br>on<br>Building | USN-5 | Deploy<br>the<br>modelin<br>the IBM<br>cloud and<br>build the<br>system | 10 | High       | Muthu Annamalai<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar.<br>Charan Kumar    |
| Sprint-4 | Training<br>and<br>testing  | USN-6 | Testing the model's perform ance and training it                        | 7  | Medi<br>um | Muthu Annamalai.V<br>KOULURU<br>NANDAKISHORE<br>REDDY<br>MukeshKumar.<br>SCharan Kumar |

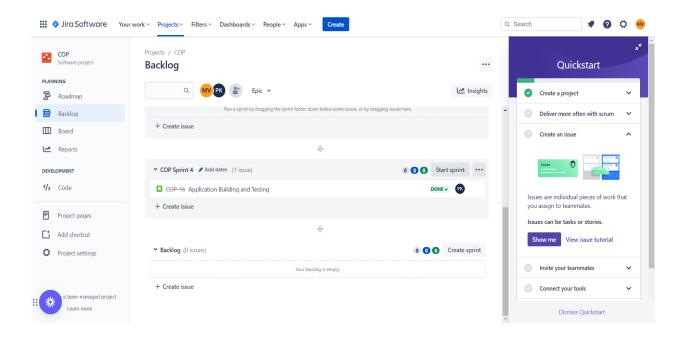
## **6.2 Sprint Delivery Schedule**

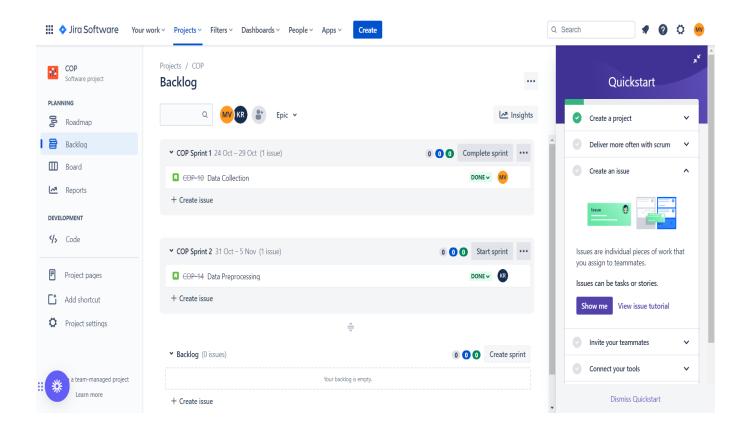
| Sprint       | Total<br>Sto<br>ry<br>Poin<br>ts | Durati<br>on | SprintSt<br>art Date | Sprint<br>End<br>Date<br>(Plan<br>ned) | Story Points Completed (as on Planned End Date) | Sprint Release Date (Actual) |
|--------------|----------------------------------|--------------|----------------------|--|---|------------------------------|
| Sprin<br>t-1 | 10                               | 6<br>Days    | 24 Oct<br>2022       | 29<br>Oct<br>2022                      | 8   | 29 Oct 2022                  |
| Sprin<br>t-2 | 10                               | 6<br>Days    | 31 Oct<br>2022       | 05<br>Nov<br>2022                      | 7   | 05 Nov 2022                  |
| Sprin<br>t-3 | 10                               | 6<br>Days    | 07 Nov<br>2022       | 12<br>Nov<br>2022                      | 8   | 12 Nov 2022                  |
| Sprin<br>t-4 | 10                               | 6<br>Days    | 14 Nov<br>2022       | 19<br>Nov<br>2022                      | 7   | 19 Nov 2022                  |

## 6.3 Reports from Jira









## 7. Coding and Solutioning

```
import Flask, render_template, request, redirect
import numpy as np
from tensorflow import keras
from keras.models import load_model
import joblib
import scipy

app = Flask(__name__)
model = load_model(r'C:\Users\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\crude-oil.h5')

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

```
@app.route('/predict',methods=["POST","GET"])
def predict():
    if request.method == "POST":
        string = request.form['val']
        string = string.split(',')
        x_input = [eval(i) for i in string]
        sc = joblib.load(r'C:\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\scaler.save')

        x_input = sc.fit_transform(np.array(x_input).reshape(-1,1))

        x_input = np.array(x_input).reshape(1,-1)

        x_input = x_input.reshape(1,-1)

        x_input = x_input.reshape(1,-1)

        x_input = x_input.reshape(1,10,1))
        print(x_input.shape)

        model = load_model(r'C:\Users\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\crude-oil.h5')
        output = model.predict(x_input)
        print(output[0][0])

        val = sc.inverse_transform(output)

        return render_template('web.html' , prediction = val[0][0])

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

if __name__=="__main__":
        app.run(debug=True)
```

# 8. Testing

## 8.1 Test cases

| Test<br>case ID | Feature<br>Type | Component    | Test<br>Scenar<br>io  | Expected<br>Result                                    | Actual<br>Result          | Status |
|-----------------|-----------------|--------------|---|---|---------------------------|--------|
| HP_TC_<br>001   | UI              | HomePa<br>ge | Verify UI elemen ts in the Home Page                                      | The Home page must be displayed properly              | Working as expected       | PASS   |
| HP_TC_<br>002   | UI              | HomePa<br>ge | Check if the UI elements are displayed properly in different screen sizes | The Home page must be displayed properly in all sizes | notdispla<br>yed          | PASS   |
| HP_TC_<br>003   | Functional      | Web Page     | Check if<br>user can<br>enter<br>thepast<br>days                          | The input price should be updated to the              | Working<br>asexpe<br>cted | PASS   |

|               |            |          | price  | applicat<br>ion<br>succes<br>sfully                                 |   |      |
|---------------|------------|----------|--|---|---|------|
| WP_TC_<br>001 | Functional | Web Page | Check if user cannot enter anynumb er as price                       | The applicati on should not allow user to enter any number as price | User is<br>able to<br>enterany<br>price | FAIL |
| WP_TC_<br>002 | Functional | Web Page | Check if the page redirects to theresult page oncethe input is given | Thepage<br>should<br>redirect<br>to the<br>results<br>page          | Working<br>asexpe<br>cted               | PASS |

## **8.2 USER ACCEPTANCE TESTING**

### **8.2.1 DEFECT ANALYSIS**

| Resolution     | Severity 1 | Severity<br>2 | Severi<br>ty 3 | Severi<br>ty 4 | Total |
|----------------|------------|---------------|----------------|----------------|-------|
| By Design      | 1          | 0             | 1              | 0              | 2     |
| Duplicate      | 0          | 0             | 0              | 0              | 0     |
| External       | 0          | 0             | 2              | 0              | 2     |
| Fixed          | 4          | 1             | 0              | 1              | 6     |
| Not Reproduced | 0          | 0             | 0              | 1              | 1     |
| Skipped        | 0          | 0             | 0              | 1              | 1     |
| Won't Fix      | 1          | 0             | 1              | 0              | 2     |
| Total          | 6          | 1             | 4              | 3              | 14    |

## **8.2.2 TESTCASE ANALYSIS**

| Section                   | Total<br>Cases | Not<br>Tested | Fail | Pa<br>ss |
|---------------------------|----------------|---------------|------|----------|
| Client<br>Applicati<br>on | 10             | 0             | 3    | 7        |
| Security                  | 2              | 0             | 1    | 1        |
| Performan<br>ce           | 3              | 0             | 1    | 2        |
| Exception<br>Reporting    | 2              | 0             | 0    | 2        |
| Print<br>Engine           | 2              | 0             | 0    | 2        |
| Final<br>ReportOut<br>put | 2              | 0             | 0    | 2        |

## 9. Results

# 9.1 Performance Metrics

| S.No. | Parameter     | Values                | Screenshot  |  |                             |  |
|-------|---------------|-----------------------|---|--|-----------------------------|--|
| 1.    | Model Summary |                       |   |  |                             |  |
|       |               |                       | Total params: 50,851 Trainable params: 50,851 Non-trainable params: 0 |  |                             |  |
|       |               |                       | dense_1 (Dense)   | (None, 1)  | 51                          |  |
|       |               |                       | lstm_5 (LSTM)   | (None, 50)   | 20200                       |  |
|       |               |                       | lstm_4 (LSTM)   | (None, 10, 50)   | 20200                       |  |
|       |               |                       | lstm_3 (LSTM)   | (None, 10, 50)   | 10400                       |  |
|       |               |                       | Layer (type)  | Output Shape   | Param #                     |  |
|       |               |                       | Model: "sequential_1"   |  |                             |  |
|       |               |                       | <pre>model.add(Dense (1)) model.summary()</pre>                       |  |                             |  |
| 2.    | Accuracy      | Training Accuracy -   |   |  |                             |  |
|       |               | 1.9685525432167308    | <pre>model.fit(x_train,y_train,valid</pre>                            | lation_data=(x_test,y_test),epochs=5                                       | 50,batch_size=64,verbose=1) |  |
|       |               |                       | Output exceeds the <u>size limit</u> . Oper                           |  |                             |  |
|       |               | Validation Accuracy - | 84/84 [<br>Epoch 2/58   | ] - 2s 26ms/step - loss: 1.2070e   |                             |  |
|       |               | 2.201959455277266     | 84/84 [<br>Epoch 3/50<br>84/84 [                                      | ] - 2s 26ms/step - loss: 1.2549e<br>] - 2s 25ms/step - loss: 1.2198e       |                             |  |
|       |               |                       | Epoch 4/58<br>84/84 [<br>Epoch 5/58                                   | ===] - 2s 25ms/step - loss: 1.3271e  | -04 - val_loss: 7.9658e-04  |  |
|       |               |                       | 84/84 [<br>Epoch 6/58<br>84/84 [                                      | ] - 2s 25ms/step - loss: 1.2178e   |                             |  |
|       |               |                       | Epoch 7/50<br>84/84 [   | ] - 2s 25ms/step - loss: 1.1613e   |                             |  |
|       |               |                       | Epoch 8/58<br>84/84 [<br>Epoch 9/58                                   | ] - 2s 25ms/step - loss: 1.0007e   | :=                          |  |
|       |               |                       | 84/84 [<br>Epoch 10/50<br>84/84 [                                     | ===] - 2s 25ms/step - loss: 1.1960e<br>===] - 2s 26ms/step - loss: 1.1780e |                             |  |
|       |               |                       | Epoch 11/50<br>84/84 [  | ] - 2s 25ms/step - loss: 1.0194e   |                             |  |
| 0     | 0 61          | Class Detected-9      |   |  |                             |  |
| 3.    | Confidence    |                       |   |  |                             |  |
|       | Score         |                       |   |  |                             |  |
|       |               | ConfidenceScore9      |   |  |                             |  |
|       |               |                       |   |  |                             |  |
|       |               |                       |   |  |                             |  |
|       |               |                       |   |  |                             |  |
|       |               |                       |   |  |                             |  |
|       |               |                       |   |  |                             |  |
|       |               |                       |   |  |                             |  |

### 10. Advantages and Disadvantages

#### **ADVANTAGES**

- 1. With Price of fallingup full to its lowestlevel in consumer will spendto gasoline in government estimate.
- 2. Tepid inflationdeclining energy price lamp down inflation.
- 3. Lowest oil prices economyenergy producing in US, Iran, Venereal.
- 4. The plungein oil price is roiling market worldwide.
- 5. Increase fuel mileage for passenger can seemed expensive

#### **DISADVANTAGES**

- 1. Falling oil prices hurt a key sector of stock market.
- 2. Less business spending automation of energy and equipment firms.
- 3. Sagging economics higher supply mayor reasonfor oil drop, investor worry persistent declines.
- 4. Less business spending automation of energy and equipment forms. These facilities high demand of actuatorand values

### 11. Conclusion

For forecasting prices, LSTM networks are better than traditional neural networks because they use back propagation models. A traditional neural network, such as an RNN, predicts the next outgoing data but does not necessarily save the previous data. RNNs are based on feed-forwarding, so the previous data is not required to predict the future data. The LSTM method stores the previous data and makes predictions based on it, which is rather encouraging and approximate. As a result, relatively encouraging results were derived. It appears that large lookups do not necessarily improve the accuracy of crude oil price predictions. Therefore, it can be concluded that the LSTM model with a single node is the most accurate.

### 12. Future Scope

This work is carried out on the closing price of crude oil; however, there are various other factorswhich also affect the crude oil prices like change in the prices and quantities (demand and supply), change in the economy and current affairs as shown by the media. The main advantage of this researchis in capturing the changingpattern of theseprices. In the coming future, fundamental indicators and market trendshave been planned to be incorporated into a model which help the proposed model perform more efficiently.

## 13. Appendix

#### **Source Code**

#### **Model Creation**

```
MODEL BUILDING:

IMPORTING THE MODEL BUILDING LIBRARIES

from tensorflow keras models import Sequential
from tensorflow keras layers import Dense
from tensorflow keras layers import LSTM

INITIALIZING THE MODEL

model = Sequential()

ADDING LSTM LAYERS

model.add(LSTM(50,return_sequences=True,input_shape=(10,1)))
model.add(LSTM(50, return_sequences=True))
model.add(LSTM(50))
```

#### Flask App

```
import Flask, render_template, request, redirect
import numpy as np
from tensorflow import keras
from keras.models import load_model
import joblib
import scipy

app = Flask(__name__)
model = load_model(r'C:\Users\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\crude-oil.h5')

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

```
@app.route('/predict',methods=["POST","GET"])
def predict():
    if request.method == "POST":
        string = request.form('val']
        string = string.split(',')
        x_input = [eval(i) for i in string]
        sc = joblib.load(r'C:\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\scaler.save')

        x_input = sc.fit_transform(np.array(x_input).reshape(-1,1))

        x_input = np.array(x_input).reshape(1,-1)

        x_input = x_input.reshape(1,-1)

        x_input = x_input.reshape(1,-1)

        x_input = x_input.reshape((1,10,1))
        print(x_input.shape)

        model = load_model(r'C:\Users\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\crude-oil.h5')
        output = model.predict(x_input)
        print(output[0][0])

        val = sc.inverse_transform(output)

        return render_template('web.html', prediction = val[0][0])

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

if __name__=="__main__":
        app.run(debug=True)
```

#### **RECOGNIZER**

```
x_input = [eval(i) for i in string]
sc = joblib.load(r'C:\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\scaler.save')

x_input = sc.fit_transform(np.array(x_input).reshape(-1,1))

x_input = np.array(x_input).reshape(1,-1)

x_input = x_input.reshape(1,-1)

x_input = x_input.reshape((1,10,1))

print(x_input.shape)

model = load_model(r'C:\Users\muthu\OneDrive\Desktop\Crude-Oil-Price-Prediction\crude-oil.h5')
output = model.predict(x_input)
print(output[0][0])

val = sc.inverse_transform(output)
```

#### Home Page - HTML

### Home Page - CSS

```
# margin: 0;
padding: 0;
}
body {
font-family: "Poppins", sans-serif;
}
.wrapper {
width: 1170px;
margin: auto;
}
header {
background: linear-gradient(□rgba(0, 0, 0, 0.8), □rgba(0, 0, 0, 0.8)),
url(2.jpy);
height: 180vh;
-webkit-background-size: cover;
background-size: cover;
background-size: cover;
background-size: cover;
background-size: neuter center;
position: relative;
}
.menu {
float: right;
list-style: none;
margin-top: 30px;
}
.menu li {
display: inline-block;
}
.menu li a {
color: □riff;
text-decoration: none;
padding: 5px 20px;
font-family: "Poppins", sans-serif;
font-size: lopx;
text-transform: uppercase;
}
```

### Web Page - HTML

Web Page - CSS

```
" {
    margin: 0;
    padding: 0;
    text-decoration: none;
    list-style: none;
}
body {
    font-family: "poppins", sans-serif;
}
.homepage {
    min-height: 100Wh;
    width: 100%;
    position: relative;
    background: linear-gradient(□rgba(0, 0, 0, 0.8), □rgba(0, 0, 0, 0.8)),
    url(1.jpg);
    background-size: cover;
    background-position: center;
    z-index: 1;
}
.homepage::after {
    content: "";
    position: absolute;
    left: 0;
    top: 0;
    height: 100%;
    width: 100%;
    background-color: □rgb(34, 33, 33);
    opacity: 0.3;
    z-index: -2;
}
.navbar {
    position: absolute;
    padding: 40px 40px 40px;
    display: flex;
    justify-content: space-between;
    width: 100%;
    box-sizing: border-box;
```

**Github Project Link:** https://github.com/IBM-EPBL/IBM-Project-6934-1658843459

### **Project Demo**

**Link:**https://drive.google.com/drive/folders/1Qlq2mmC1XJ\_7slnsAuRoW5 yLOIM6sR0N?usp=sharing