CRUDE OIL PREDICTION

PROJECT REPORT

Team Id: PNT2022TMID00817

Submitted by

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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 Project 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
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- 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)
- 5. Alquist, R., Kilian, L., Vigfusson, R.J., 2013. Forecasting the price of oil. In: Elliott G
- 6. Bifet, A., Kirkby, R., Kranen, P., Reutemann, P., 2012. Massive online analysis manual
- 7. N. Raj Kiran, V. Ravi. "Software reliability prediction by soft computing techniques", Journal of Systems and Software, 2008
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- 9. Crude Oil Spot Price Forecasting", Studies in Fuzziness and Soft Computing, 2008

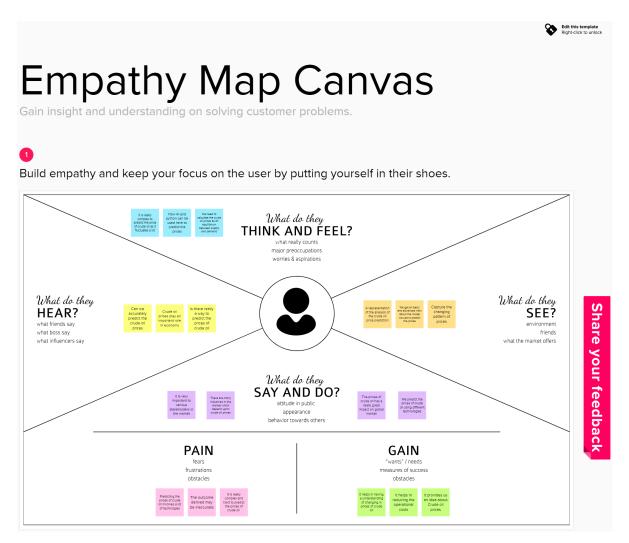
2.3 Problem Statement Definition

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 and Proposed Solution

3.1 Empathy Map Canvas

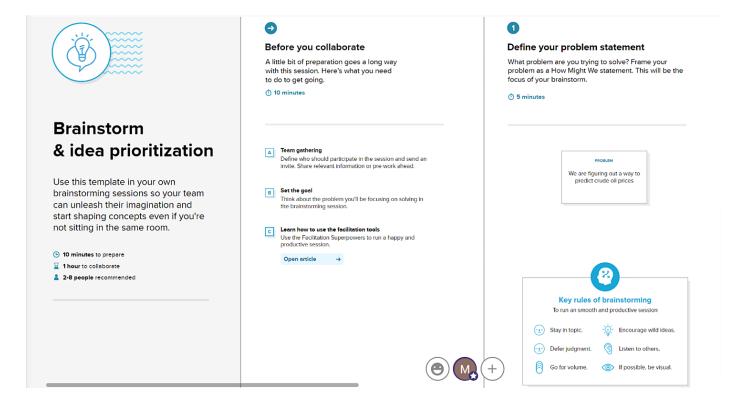
Empathy maps are simple easy-to-digest visuals that convey knowledge about a user's attitude and behavior. Team members can use it to better understand their users with the help of this tool. Understanding the true problem and the person experiencing it is essential for creating an effective solution. 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.



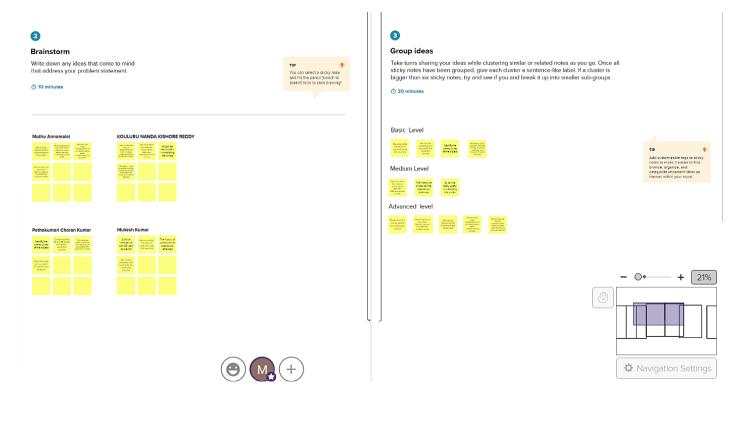
3.2 Ideation and Brainstorming

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.

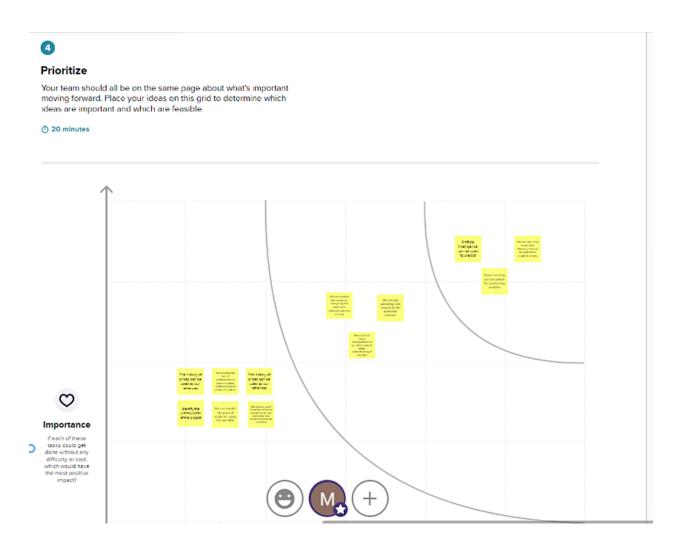
Step 1: Team Gathering, Collaboration and Select the Problem Statement



Step 2: Brainstorm, Idea Listing and Grouping



Step 3: Idea Prioritization



3.3 Proposed Solution

S.N	Parameter	Description
Ο.		
1.	Problem Statement (Problem to be 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	The crude oil price can be easilypredicted by
	description	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/ Customer	Oil price changes have important consequences
	Satisfaction	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.		Our revenuemodel is focused on
	Model)	 Pay per month model Pay per year model

6.	Scalability of the Solution	The time series analysis method is used to predict
		crude oil prices on the basis of previous historica
		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	Functional	Sub Requirement (Story/ Sub-Task)			
No.	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/Mobile number			

4.2 Non-Functional Requirements

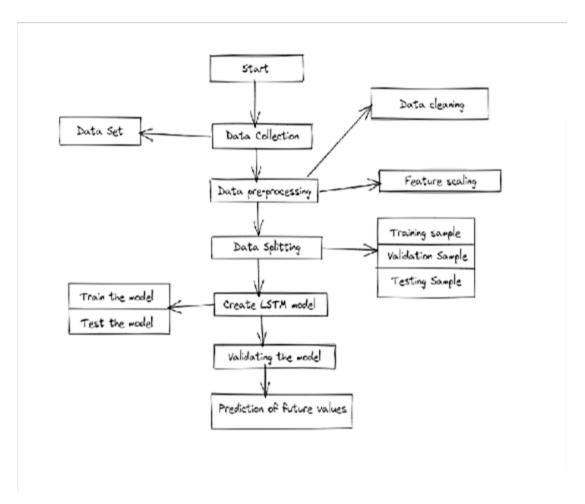
Following are the non-functional requirements of the proposed solution.

FR	Non-Functional	Description
No.	Requirement	
NFR-	Usability	UI is user-friendly .we represent
1		the data in chartswhichuses clear
		understanding of price activity
NFR-	Security	We follow certain security protocols
2		likeusing user
		credentials , OTP verification
NFR-	Reliability	The Data which represented in
3		web app is soaccurate 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

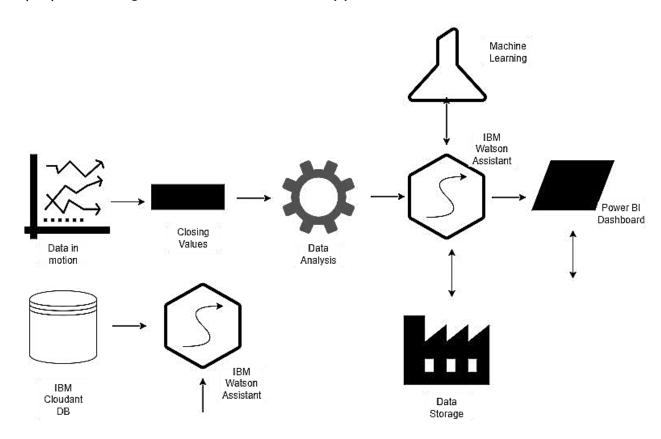
Data Flow Diagrams(DFDs) are traditional visual representations of how information flows within a system. It is possible to illustrate the right amount of the system requirement graphically with a neat and clear DFD. It explains how data enters and leaves the system, what changes the information, and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



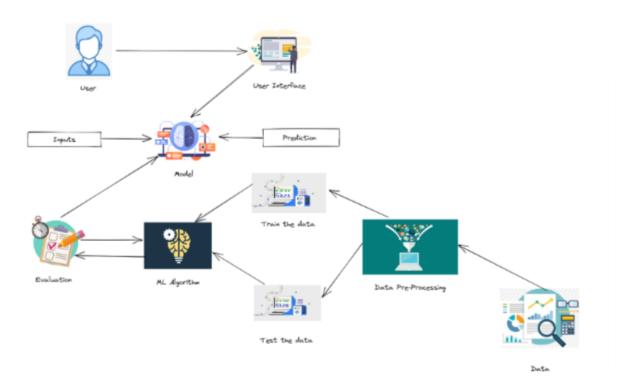
5.2 Solution and Technical Architecture

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). Solutionarchitecture 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 Requirement (Epic)	User Story Numb er	User Story / Task
Sprint-1	Data Collection	USN-1	Collecting the Dataset
	Data Pre- processing	USN-2	Data Pre-processing
Sprint-3	Model Building	USN-3	Prepare the model by importing thenecessary libraries, adding the layers, and compiling it.
Sprint-3	Model Building	USN-4	The data classification model istrained using RNNs and other systems.

Sprint-4	Application 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 Schedule

6.1 Sprint Planning and Estimation

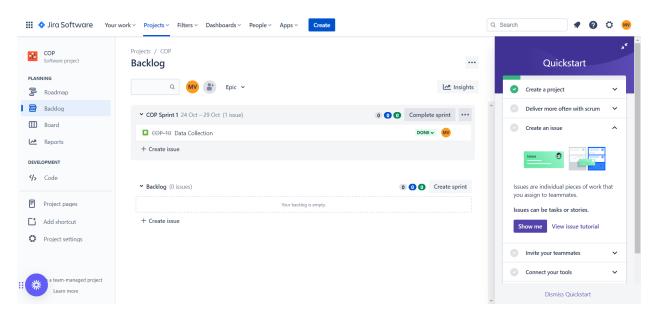
Sprint	Functional Requireme nt (Epic)	User Story Numb er	User Story / Task	Story Poin ts	Priority	Team Members
Sprint-1	Data	USN-1	Collecting the	10	High	Muthu Annamalai.V
	Collection		Dataset			KOULURU NANDA
						KISHORE REDDY
						MukeshKumar. S Charan Kumar
						Charan Kumai
Sprint-2	Data Pre-	USN-2	Data Pre-	7	Medi	Muthu Annamalai.V
	processing		processing		um	KOULURU NANDA
						KISHORE REDDY
						MukeshKumar. S Charan Kumar
						Charan Kumai
Sprint-3	Model	USN-3	Prepare the	10	High	Muthu Annamalai.V
	Building		model by importing			KOULURU NANDA
			thenecessary			KISHORE REDDY
			libraries,			MukeshKumar. S Charan Kumar
			adding the			Charan Kumar
			layers, and			
Sprint-3	Model	USN-4	compiling it. The data	7	Medi	Muthu Annamalai.V
	Building		classificati		um	KOULURU NANDA
			on model is			KISHORE REDDY
			trained using RNNs			MukeshKumar. S
			and other			Charan Kumar
			systems.			

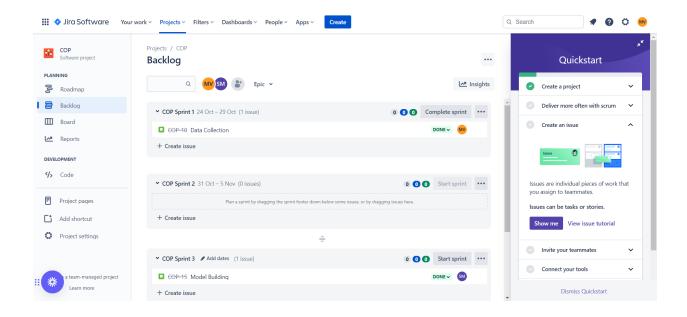
Sprint-4	Application Building	Deploy the modelin the IBMcloudand build the system	10		Muthu Annamalai.V KOULURU NANDA KISHORE REDDY MukeshKumar. S Charan Kumar
Sprint-4	Training and testing	Testing the model's performancea nd training it	7	um	Muthu Annamalai.V KOULURU NANDA KISHORE REDDY MukeshKumar. S Charan Kumar

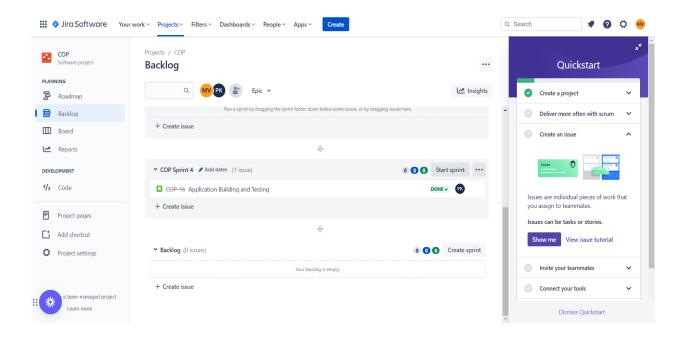
6.2 Sprint Delivery Schedule

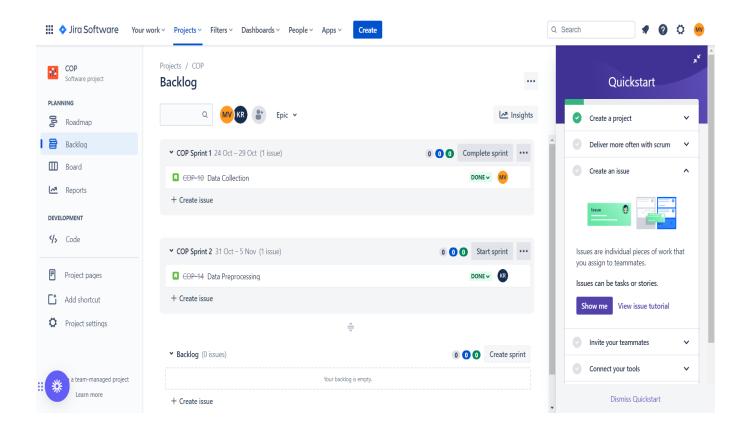
Sprint	Total Story Points	Durati on	SprintStart Date	Sprint End Date (Planne d)	Story Points Completed (as on Planned End Date)	Sprint Release Date(Actual)
Sprint-1	10	6 Days	24 Oct 2022	29 Oct 2022	8	29 Oct 2022
Sprint-2	10	6 Days	31 Oct 2022	05 Nov 2022	7	05 Nov 2022
Sprint-3	10	6 Days	07 Nov 2022	12 Nov 2022	8	12 Nov 2022
Sprint-4	10	6 Days	14 Nov 2022	19 Nov 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"])
idef home():
    return render_template('index.html')
```

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

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

8. Testing

8.1 Test cases

Test case ID	FeatureType	Component	Test Scenario	Expected Result	Actual Result	Status
HP_TC_001	UI	HomePage	Verify UI elements in the Home Page	The Home page must be displayed properly	Working asexpect ed	PASS
HP_TC_002	UI	HomePage	Check if the UI elements are displayed properly in different screen sizes	The Home page must be displayed properly in all sizes	The UI is notdisplayed properly in screen size 2560 x 1801 and 768 x 630	PASS
HP_TC_003	Functional	Web Page	Check if user can enter thepast days price	The input price shouldbe updated to the application successfully	Working asexpect ed	PASS
WP_TC_001	Functional	Web Page	Check if user cannot enter anynumber as price	The application should not allow user to enter any number as price	User is able to enterany price	FAIL
WP_TC_002	Functional	Web Page	Check if the page redirects to theresult page oncethe input is given	Thepage should redirect to the results page	Working asexpect ed	PASS

8.2 USER ACCEPTANCE TESTING

8.2.1 DEFECTANALYSIS

Resolution	Severity 1	Severity 2	Severity 3	Severity 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
NotReproduced	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	TotalCases	Not Tested	Fail	Pass
Client Application	10	0	3	7
Security	2	0	1	1
Performance	3	0	1	2
Exception Reporting	2	0	0	2
Print Engine	2	0	0	2
Final ReportOutput	2	0	0	2

Model Performance Testing:

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	
			Model: "sequential_1"		
			<pre>model.add(Dense (1)) model.summary()</pre>		
2.	Accuracy	Training Accuracy -			
		1.9685525432167308	<pre>model.fit(x_train,y_train,validat</pre>	ion_data=(x_test,y_test),epochs=50,ba	atch_size=64,verbose=1)
			Output exceeds the <u>size limit</u> . Open Epoch 1/50	the full output data <u>in a text editor</u>	
		Validation Accuracy –	84/84 [Epoch 2/50	-] - 2s 26ms/step - loss: 1.2070e-04	
		2.201959455277266	Epoch 3/50 84/84 [-] - 2s 26ms/step - loss: 1.2549e-04 -] - 2s 25ms/step - loss: 1.2190e-04	
			Epoch 4/58 84/84 [====== Epoch 5/58	=] - 2s 25ms/step - loss: 1.3271e-04	- val_loss: 7.9658e-04
			Epoch 6/50	=] - 2s 25ms/step - loss: 1.2178e-84 =] - 2s 25ms/step - loss: 1.1164e-84	22
			Epoch 7/50 84/84 [Epoch 8/50	-] - 2s 25ms/step - loss: 1.1613e-84	- val_loss: 7.4860e-04
			84/84 [Epoch 9/50	-] - 2s 25ms/step - loss: 1.0007e-04 -] - 2s 25ms/step - loss: 1.1960e-04	
			Epoch 10/50 84/84 [=] - 2s 26ms/step - loss: 1.1788e-84	
			Epoch 11/50 84/84 [=] - 2s 25ms/step - loss: 1.0194e-04	- val_loss: 6.4463e-04
3.	Confidence	Class Detected-9			
0.	Score				
	Score	ConfidenceScore9			
		Confidencescores			

10. Advantages and Disadvantages

ADVANTAGES

- 1. With Price of fallingup full to its lowestlevel in consumerwill 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 roilingmarket worldwide.
- 5. Increase fuel mileage for passenger can seemed expensive

DISADVANTAGES

- 1. Falling oil prices hurt a key sector of stock market.
- 2. Less businessspending automation of energy and equipment firms.
- 3. Sagging economics higher supply mayor reasonfor oil drop, investorworrypersistent declines.
- 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

- 1. This work is carried out on the closing price of crude oil; however, there arevarious other factors which also affect the crude oil prices like change in theprices and quantities (demand and supply), change in the economy and current affairs as shown by the media.
- 2. The main advantage of this researchis in capturing the changing pattern of these prices.
- In the coming future, fundamental indicators and market trendshave beenplanned to be incorporated into a model which help the proposed model perform more efficiently.

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"])
idef 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(n'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(n'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

Web Page - HTML

Web Page - CSS

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

Project Demo Link:

https://drive.google.com/drive/folders/1Qlq2mmC1XJ_7slnsAuRoW5yLOIM6sR0N?usp=sharing