

# **CRUDE OIL PRICE PREDICTION PROJECT REPORT**

**TEAM ID : PNT2022TMID43614**

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# **1.INTRODUCTION**

## **1.1 PROJECT OVERVIEW:**

Crude oil is the world's leading fuel, and its prices have a big impact on the global environment, economy as well as oil exploration and exploitation activities. Oil price forecasts are very useful to industries, governments and individuals. Although many methods have been developed for predicting oil prices, it remains one of the most challenging forecasting problems due to the high volatility of oil prices. In this paper, we propose a novel approach for crude oil price prediction based on a new machine learning paradigm called stream learning. The main advantage of our stream learning approach is that the prediction model can capture the changing pattern of oil prices since the model is continuously updated whenever new oil price data are available, with very small constant overhead. To evaluate the forecasting ability of our streaming learning model, we compare it with three other popular oil price prediction models. The experiment results show that our stream learning model achieves the highest accuracy in terms of both mean squared prediction error and directional accuracy ratio over a variety of forecast time horizons.

## **1.2 PURPOSE:**

Crude oil price fluctuations have a far reaching impact on global economies and thus price forecasting can assist in minimising the risks associated with volatility in oil prices. Price forecasts are very important to various stakeholders: governments, public and private enterprises, policymakers, and investors.

## **2.LITERATURE SURVAY**

### **2.1 Existing problem:**

The prediction of the crude oil rates based on the previous datasets on the data and prices as the feature \_list are inputs and target list are predicted values. The implementation was on the Linear Regression Model which is feasible to some extent for the prediction of the crude oil prices. The implementation is on predicting the crude oil prices for the days using Linear Regression Python Machine Algorithm and plotting the graph based on the prediction.

### **2.2 Reference:**

According to economic theory, the price of crude oil should be easily predictable from the equilibrium between demand and supply, wherein demand forecasts are usually made from GDP, exchange rates and domestic prices, and supply is predicted from past production data and reserve data.

### **2.3 Problem statement definition:**

There are five main problems identified based on investigations made on previous research. Firstly, data used in the previous predictions are majority employed from WTI or Brent crude oil price without taking into consideration other inputs that are involved together in the market. The crude oil price market volatiles from the contributions made by other factors surround it and neglecting these factors will demote the capability of a prediction tool. A good prediction is the one that can

comprehends and correlates between factors, sparks information on the trend and finally, predict it accurately.

Secondly, there are scarce numbers of research that implement the verification and validation technique on the main factors involving in the fluctuation. Besides the global crude oil price, other popular factors that being used in previous research are demand and supply. Although, demand and supply of oil plays vital role to the market volatility, the use of these observations only is not enough to comprehensively render the information offered by the trend.

There are also other factors that contributed to the trend and gave impact to the price. Therefore, by embracing appropriate key factors and later correlate them will help to achieve a thorough and comprehensive prediction for the market.

Thirdly, time-series data are mainly used for prediction.

Nevertheless, data pre-processing and data representation process are made absent in some of the previous research.

These two processes are important to cleanse and reduce errors and noises in data set and uniform it. Later, these will help to organise the process of prediction, make it more systematic and finally, generates more stable result. Without these processes, the prediction tool will be less reliable.

Fourthly, the crude oil price movement was the popular topic studied previously and not the crude oil price itself.

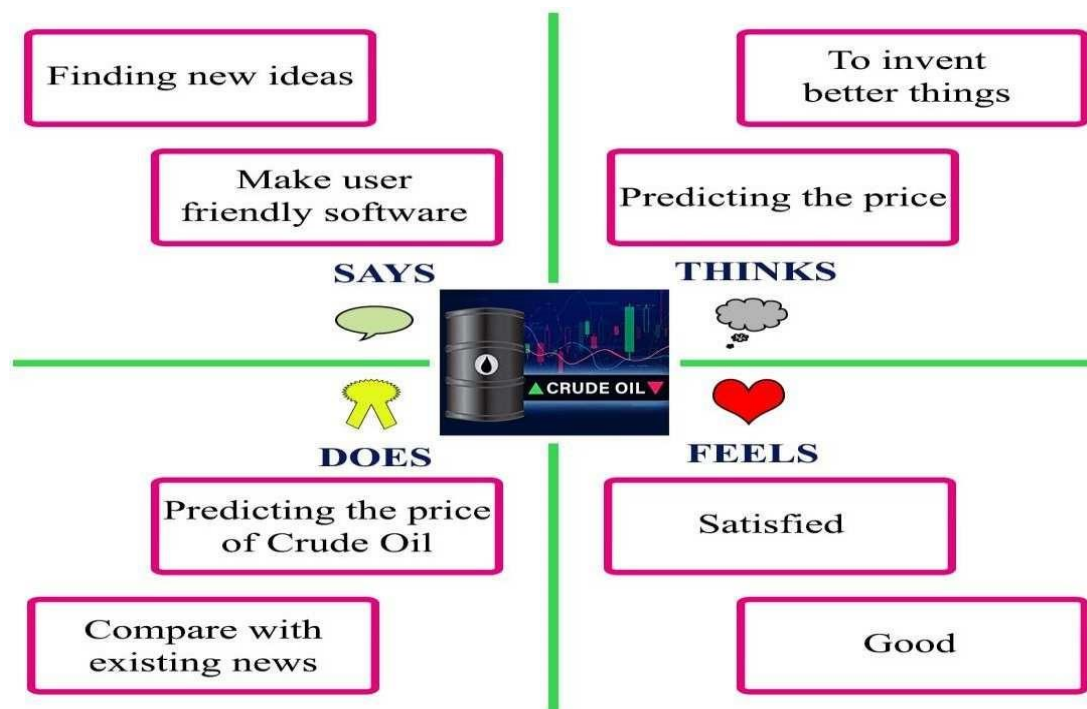
Predicting the movement of the price only is not sufficient to characterise the market where else, crisp prediction will offer far more persona. A prediction on the movement together with the price itself will tender more usable, discrete and practical implementation to the real world problem.

Sincerely, the practicability of the previous study is still dubious as the crude oil market itself is chaotic. Still, there are opportunities for

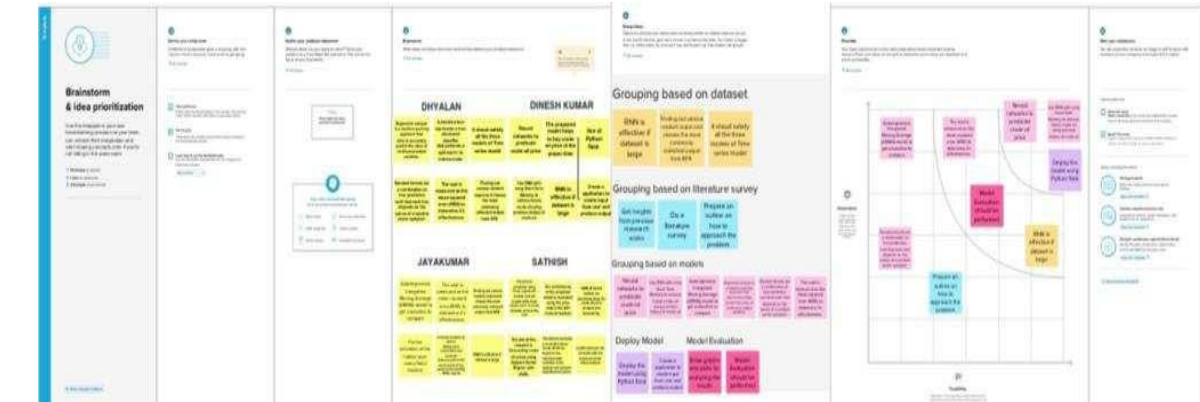
improvement in the future as the advancement of our world technology is rapid.

## **3.IDEATION AND PROPOSED SOLUTION**

### **3.1 Empathy map and canvas**



## 3.2 Ideation & Brainstrom





### **3.3 Proposed solution**

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	The price of crude oil, the most essential fuel in the world, has a big impact on the environment globally, thus forecasts are very helpful for governments, businesses, and people. Continuous use of statistical and econometric methods, including AI, may have a negative impact on prediction accuracy.
2.	Idea / Solution description	In order to predict future crude oil using historical data on crude oil, RNN is utilised with long short-term memory. The effectiveness of the cost is calculated using the mean squared error. Using the pricing information in the WTO crude oil materials, the proposed model's performance is assessed.
3.	Novelty / Uniqueness	<ul style="list-style-type: none"><li>● Crude oil price variations have a significant impact on the world's economies, thus price forecasting can help reduce the risks brought on by this volatility.</li><li>● For a variety of stakeholders, including governments, public and private businesses, legislators, and investors, price projections are crucial.</li></ul>
4.	Social Impact / Customer Satisfaction	<ul style="list-style-type: none"><li>● It is employed to forecast future pricing and consume oil in accordance with such prices.</li><li>● This price directly affects a number of goods and products, and its changes have an impact on the stock markets.</li><li>● In addition to economic factors, significant events can have an impact on oil prices.</li></ul>
5.	Business Model (Revenue Model)	<ul style="list-style-type: none"><li>● When deciding whether to purchase or sell crude oil, it can be useful to decision-makers who may be businesses, individual investors, or both.</li><li>● One of the most profitable commodities for traders to trade is crude oil.</li><li>● To anticipate the price of crude oil, RNN and LSTM models are employed as the benchmark model.</li></ul>
6.	Scalability of the Solution	<ul style="list-style-type: none"><li>● The dimensions of the data are reduced using the PCA, MDS, and LLE methods.</li><li>● RNN and LSTM model accuracy should be increased.</li></ul>

## 3.4 Problem solution fit

Project Title: Crude oil price prediction

Project Design Phase-I - Solution Fit Template

Team ID: PNT2022TMID43580

Define CS, fit into CC	<b>1. CUSTOMER SEGMENT(S)</b> Who is your customer? Oil accounts are the third of the world's energy consumption that is the greatest share for all category of government.	<b>6. CUSTOMER CONSTRAINTS</b> What constraints prevent your customers from taking action or limit their choices of solutions? Due to strong chain effects owned by this crude oil market, Bias in the factors involved will have exclusive impact to the price.	<b>5. AVAILABLE SOLUTIONS</b> Which solutions are available to the customer when they face the problem or need to get the job done? What have they tried in the past? What pros & cons do these solutions have? There are innumerable ways and approaches which are being used and have been used for predicting the prices of crude oil.	Explore A.S, differentiate
	<b>2. JOBS-TO-BE-DONE / PROBLEMS</b> Economic growth is one of the biggest factors affecting petroleum product—and therefore crude oil demand. Growing economies increase demand for energy in general and especially for transportation.	<b>9. PROBLEM ROOT CAUSE</b> There is only one dependent variable, the closing price of crude oil which has been considered as a time series.	<b>7. BEHAVIOUR</b> The correct information should be given by the individual.	
Focus on A.S, fit into BE, into CH	<b>3. TRIGGERS</b> What triggers customers to act? By seeing our friends and colleagues benefited by this web.	<b>10. YOUR SOLUTION</b> A contemporary and innovative method of predicting crude oil prices using the artificial neural network.	<b>8. CHANNELS of BEHAVIOUR</b> S.1 ONLINE Customer has used this web in any time any where. S.2 OFFLINE Non-working days are not predicting the price of crude oil.	CH
<b>4. EMOTIONS: BEFORE / AFTER</b> How do customers feel when they face a problem or a job and afterwards? Traveling for job will be reduced.	<b>EM</b>			

# 4.REQUIREMENT ANALYSIS

Project Title: Crude oil price prediction

Project Design Phase-I - Solution Fit Template

Team ID: PNT2022TMID43580

Define CS, fit into CC	<b>1. CUSTOMER SEGMENT(S)</b> Who is your customer? Oil accounts are the third of the world's energy consumption that is the greatest share for all category of government	<b>6. CUSTOMER CONSTRAINTS</b> What constraints prevent your customers from taking action or limit their choices of solutions? Due to strong chain effects owned by this crude oil market, bias in the factors involved will have exclusive impact to the price	<b>5. AVAILABLE SOLUTIONS</b> Which solutions are available to the customers when they face the problem or need to get the job done? What have they tried in the past? What pros & cons do these solutions have? There are innumerable ways and approaches which are being used and have been used for predicting the prices of crude oil	Explore AS, differentiate
	<b>2. JOBS-TO-BE-DONE / PROBLEMS</b> Economic growth is one of the biggest factors affecting petroleum product—and therefore crude oil demand. Growing economies increase demand for energy in general and especially for transportation.	<b>9. PROBLEM ROOT CAUSE</b> There is only one dependent variable, the closing price of crude oil which has been considered its a time series	<b>7. BEHAVIOUR</b> The correct information should be given by the individual.	
Focus on AS, fit into BE, understand	<b>3. TRIGGERS</b> What triggers customers to act? By seeing our friends and colleagues benefited by this web.	<b>10. YOUR SOLUTION</b> A contemporary and innovative method of predicting crude oil prices using the artificial neural network.	<b>8. CHANNELS of BEHAVIOUR</b> 8.1 ONLINE Customer has used this web in any time any where 8.2 OFFLINE Non-working days are not predicting the price of crude oil.	Focus on AS, fit into BE, understand
<b>4. EMOTIONS: BEFORE / AFTER</b> How do customers feel when they face a problem or a job and afterwards? Travelling for job will be reduced.	<b>8. CHANNELS of BEHAVIOUR</b> 8.1 ONLINE Customer has used this web in any time any where 8.2 OFFLINE Non-working days are not predicting the price of crude oil.			

## 4.1 Functional Requirements

- **User Application :** The price of crude oil can be downloaded directly from the Google Play Store application by users.
- **User Products Available :** The Application's User There are numerous products in the Crude Oil Price App, and users can instantly update the energy and oil prices.
- **User Additional Features :** The user can view oil price charts and the most recent news. Major Energy Quotes User View The user may use various color schemes.
- **User Exceptions :** The User Can Exchange Rates and Currency Converter.

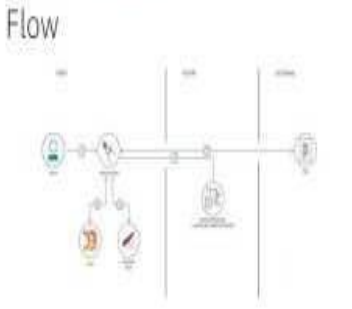
## **4.2 Non-Functional Requirements**

- Usability : Give the accurate price prediction of crude oil.
- Security : The data of user will be secure and encrypted.
- Reliability : Give the best and accurate price and knowledge to the
- Availability : The application will be available on all the platform.
- Scalability : The scalability will be approximately 95%.

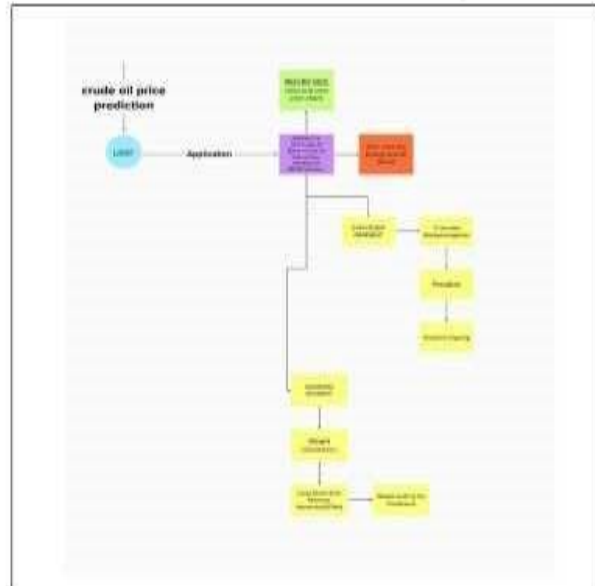
# **5.PROJECT DESIGN**

## **5.1 Data Flow Diagrams**

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1. User configures credentials for the Watson Natural Language Understanding service and starts the app.
2. User selects data file to process and load.
3. Apache Tika extracts text from the data file.
4. Extracted text is passed to Watson NLU for enrichment.
5. Enriched data is visualized in the UI using the D3.js library.

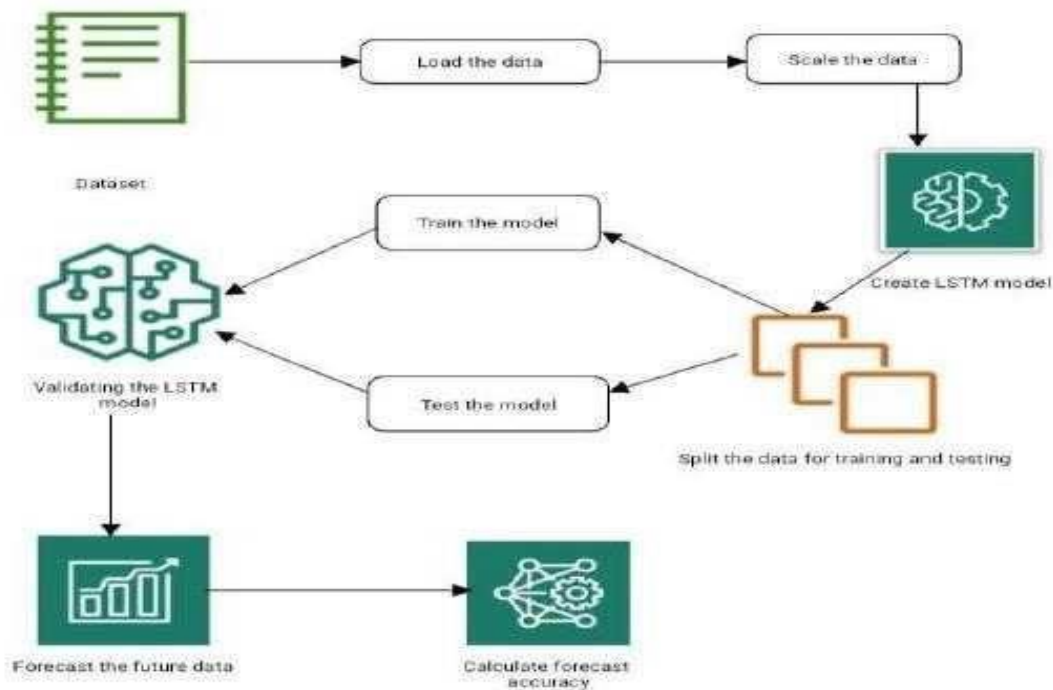


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- Find the best tech solution to solve existing problems in the organizations – LSTM technique.

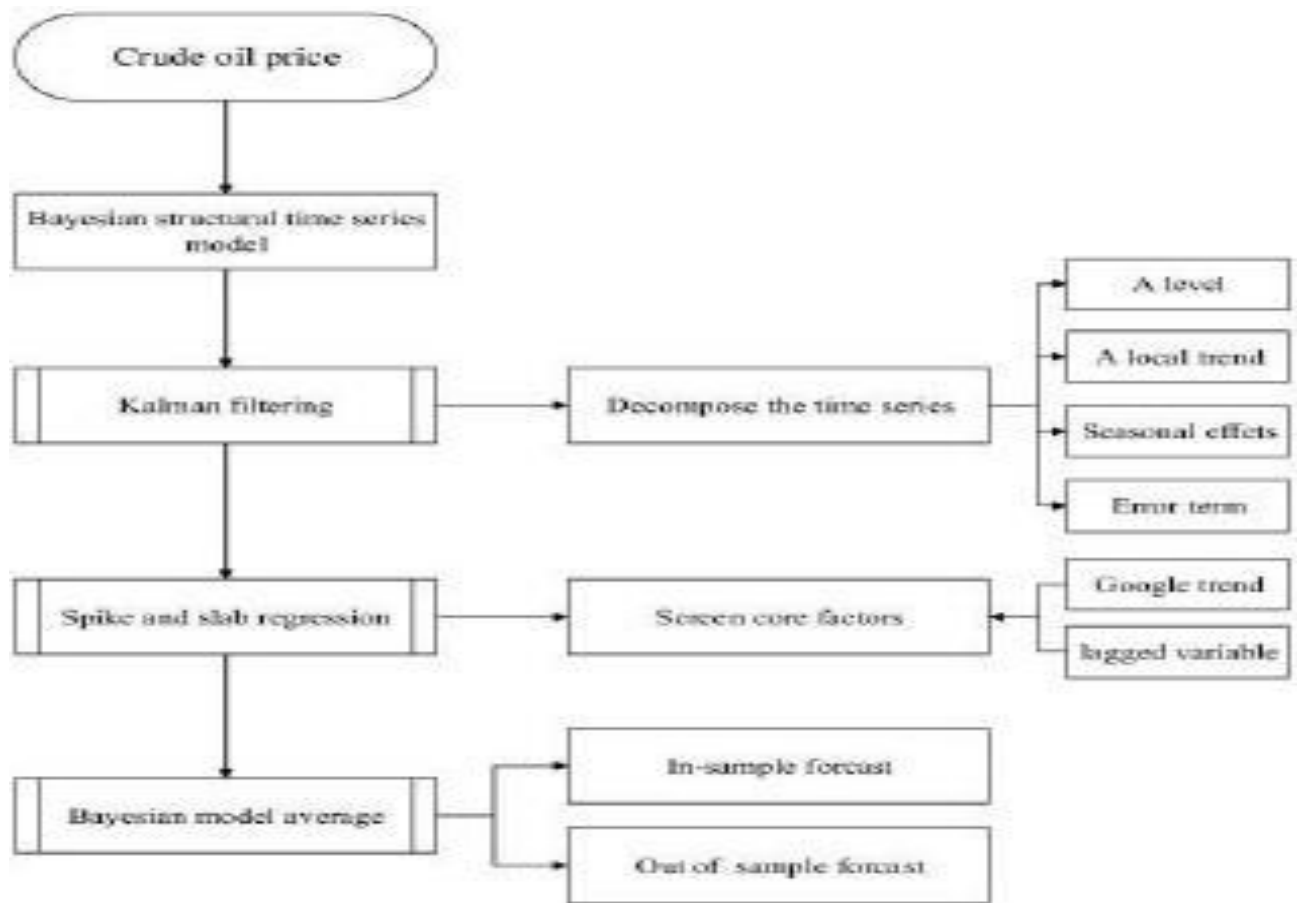
- Describe the structure, characteristics, behavior, and other aspects of the software is move to LSTM for predicting more accurate Crude Oil Prices.
- Define features, development phases, and solution requirements of the project.
- Provide specifications according to which the solution is defined, managed, and delivered.

### **.Solution Architecture Diagram:**



*Figure 1: Architecture and data flow of the Crude oil Price Prediction*

### **.Technical Architecture Diagram :**



## 5.3 User Stories

## User Stories

Use the below template to list all the user stories for the product.

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Application	USN-1	You can download the crude oil price by opening the Google Play Store app directly as a user.	I can access own decisions.	High	Sprint-1
	Available Products	USN-2	Users of the application may instantly update the energy and oil prices while using it because there are so many different products in the crude oil price app.	I can receive the data once click then confirm	High	Sprint-1
	Additional Features	USN-3	Users can read the most recent news and see oil price charts. Major Energy Quotes User View The user may use many colour schemes.	I can view then read the price prediction.	High	Sprint-2
	Expectations	USN-4	User Can Convert Currency And Exchange Rates	I can expect	Medium	Sprint-1
	Login	USN-5	Log in as a user without using an email address, username, or password.		High	Sprint-1
Customer (Web user)			As a user I can view the crude oil price	I can view the price directly	High	Sprint-1
Customer Care Executive			As a user I executive the given price history	I can accept the terms	medium	Sprint-1
Administrator			As a manager, it anticipates the results.	Show the result	High	Sprint-1

## 6.PROJECTPLANNING&SCHEDULG



## 6.1 Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-2	Registration	USN-3	As a user, I can register for the application by entering my email, password, and confirming my password.	10	High	SUBIN.T
Sprint-2		USN-4	As a user, I will receive confirmation email once I have registered for the application	10	High	HARIPRASAD.P
Sprint-2	Login	USN-5	As a user, I can log into the application by entering email & password.	15	High	SUBIN.T
Sprint-1	Input Necessary Details	USN-1	As a user, I can give Input Details to Predict Likelihood of crude oil	15	High	AFZAL.A
Sprint-1	Data Pre-processing	USN-2	Transform raw data into suitable format for prediction.	15	High	NANJUNDESWARAN
Sprint-3	Prediction of Crude Oil Price	USN-6	As a user, I can predict Crude oil using machine learning model.	20	High	HARIPRASAD.P
Sprint-3		USN-7	As a user, I can get accurate prediction of crude oil	5	Medium	NANJUNDESWARAN
Sprint-4	Review	USN-8	As a user, I can give feedback of the application.	20	High	AFZAL.A

## 6.2 Sprint Delivery Schedule

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022		
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022		

Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022		

## 6.3 Reports From JIRA

←

→

↺

pn20222bmid13112.atlassian.net/jira/software/projects/COPP/boards/1/backlog

🔍

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⚙️

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🔔

?

⚙️

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🏠

Crude Oil Price Predict...

Software project

PLANNING

📅 Roadmap

**📋 Backlog**

📊 Board

DEVELOPMENT

🔧 Code

📄 Project pages

🔖 Add shortcut

⚙️ Project settings

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Projects / Crude Oil Price Prediction

**Backlog**

...

🔍

👤

👥

Epic

📈 Insights

+ Create issue

▼ Backlog (5 issues)

📅 0 0 0 Create sprint

🟢 COPP-1

As a user, I want to collect the dataset so that I can perform price prediction with the data...

🕒 TO DO

👤

🟢 COPP-2

As a user, I want to perform data preprocessing so that my data is free of redundancies, n...

🕒 TO DO

👤

🟢 COPP-3

As a user, I want to build the model so that the model can be configured by adding LST...

🕒 TO DO

👤

🟢 COPP-4

As a user, I want to train the model so that the model can predict results with utmost ac...

🕒 TO DO

👤

🟢 COPP-5

As a user, I want to develop an application so that the predicted results can be visualized

🕒 TO DO

👤

🟢 COPP-6

As a user, I want to deploy the model on the IBM cloud.

🕒 TO DO

👤

+ Create issue

🎬

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🔍 Find help

▼

🗣️ Give feedback

Dismiss Quickstart

Crude Oil Price Predict...  
Software project

PLANNING

- Roadmap
- Backlog
- Board

DEVELOPMENT

- Code

Project pages

- Add shortcut
- Project settings

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5 days remaining Complete sprint

IN PROGRESS 1 ISSUE

COPP-2

DATA COLLECTION

COPP-3

Insights COPP SPRINT 1

DEVELOPMENT

- Code

Project pages

- Add shortcut
- Project settings

1 Oct - 5 Nov (2 issues)

+ Create issue

COPP-3 As a user, I want to build the model so that the model can be configured by adding LSTM and output layers MODEL BUILDING

5 IN PROGRESS

+ Create issue

## **7.CODING AND SOLUTIONING**

**(explain the features add in the project along with code )**

### **7.1 Feature 1**

- ❖ The crude oil prediction website provides two options
  - Home
  - predict
- ❖ The home allows the user to have an insight on the importance of crude oil price prediction
- ❖ The predict allows the user to give the 10 days input and arrive at the prediction results

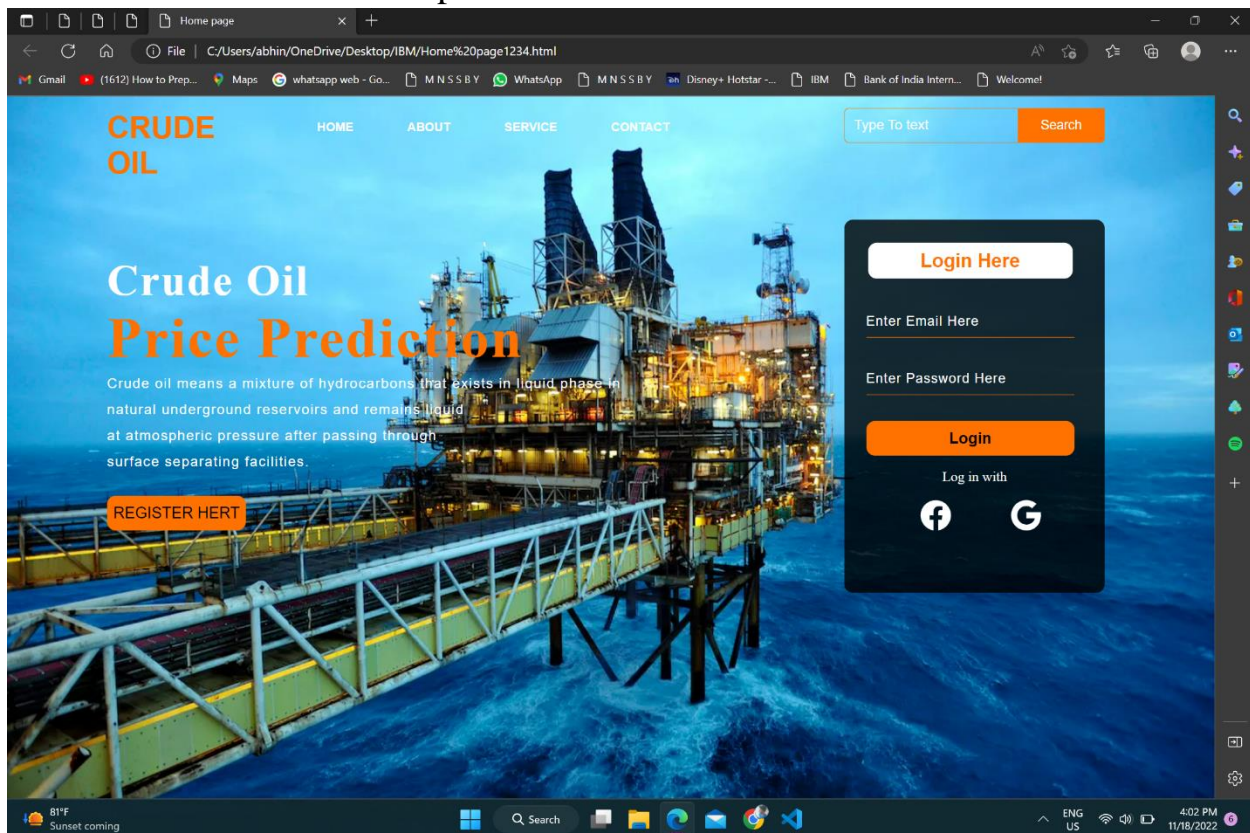


Figure 1

The screenshot shows a web browser window with a registration form titled "Register Here". The form is centered on the screen and has a light blue background. The background image of the browser is a sunset over the ocean with an offshore oil rig. The form contains the following fields and controls:

- First Name :** A text input field with the placeholder text "Enter Your First Name".
- Last Name :** A text input field with the placeholder text "Enter Your last Name".
- Your Age :** A text input field with the placeholder text "How Old Are You".
- Email :** A text input field with the placeholder text "Enter Your Valid Email".
- password:** A text input field with the placeholder text "Enter Password Here".
- Gender :** Two radio button options: "Male" and "Female".
- Submit** : An orange button at the bottom of the form.

The browser's address bar shows the file path "C:/Users/abhin/OneDrive/Desktop/IBM/register.html". The Windows taskbar at the bottom shows the date and time as "4:05 PM 11/18/2022" and the language as "ENG US".

Figure 2

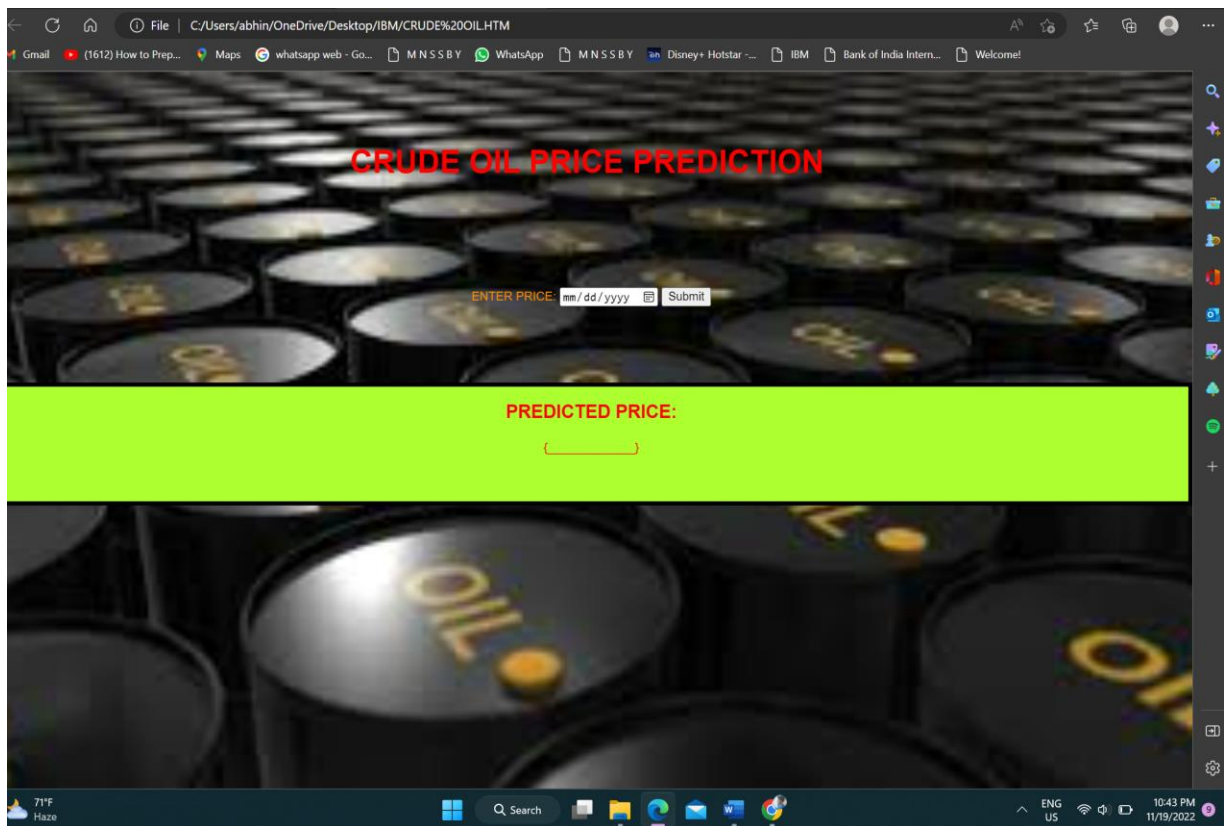


Figure 3



## CRUDE OIL PRICE PREDICTOR

Enter previous 10th day price

Enter previous 9th day price

Enter previous 8th day price

Enter previous 7th day price

Enter previous 6th day price

Enter previous 6th day price

Enter previous 5th day price

Enter previous 4th day price

Enter previous 3th day price

Enter previous 2nd day price

Enter previous 1st day price

Predict

## CODE

### Home Page:

.HTML

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
```

```

<title>Home page</title>
<link rel="stylesheet" href="style.css">
</head>
<body>

    <div class="main">
        <div class="navbar">
            <div class="icon">
                <h2 class="logo">CRUDE OIL</h2>
            </div>

            <div class="menu">
                <ul>
                    <li><a href="register.css">HOME</a></li>
                    <li><a href="register.css">ABOUT</a></li>
                    <li><a href="register.css">SERVICE</a></li>
                    <li><a href="register.css">CONTACT</a></li>
                </ul>
            </div>

            <div class="search">
                <input class="srch" type="search" name="" placeholder="Type To
text">

                <a href="#"> <button class="btn">Search</button></a>
            </div>

        </div>
        <div class="content">
            <h1>Crude Oil<br><span>Price Prediction</span><br></h1>
            <p class="par"> Crude oil means a mixture of hydrocarbons that exists
in liquid phase in<br>
                natural underground reservoirs and remains liquid <br>at
atmospheric pressure
                after passing through <br>surface separating facilities.</p>

            <button class="cn"><a href="register.html">REGISTER
HERT</a></button>

            <div class="form">
                <h2>Login Here</h2>
                <input type="email" name="email" placeholder="Enter Email
Here">

                <input type="password" name="" placeholder="Enter Password
Here">

                <button class="bttn"><a href="#">Login</a></button>

```





```
font-size: 35px;
font-family: Arial;
padding-left: 20px;
float: left;
padding-top: 10px;
margin-top: 5px
}
.menu{
width: 400px;
float: left;
height: 70px;
}
ul{
float: left;
display: flex;
justify-content: center;
align-items: center;
}
ul li{
list-style: none;
margin-left: 62px;
margin-top: 27px;
font-size: 14px;
}
ul li a{
text-decoration: none;
color: #fff;
font-family: Arial;
font-weight: bold;
transition: 0.4s ease-in-out;
}
ul li a:hover{
color: #ff7200;
}
.search{
width: 330px;
float: left;
margin-left: 270px;
}
.srch{
font-family: 'Times New Roman';
width: 200px;
height: 40px;
background: transparent;
border: 1px solid #ff7200;
```

```
margin-top: 13px;
color: #fff;
border-right: none;
font-size: 16px;
float: left;
padding: 10px;
border-bottom-left-radius: 5px;
border-top-left-radius: 5px;
}
.btn{
width: 100px;
height: 40px;
background: #ff7200;
border: 2px solid #ff7200;
margin-top: 13px;
color: #fff;
font-size: 15px;
border-bottom-right-radius: 5px;
border-bottom-right-radius: 5px;
transition: 0.2s ease;
cursor: pointer;
}
.btn:hover{
color: white;
}
.btn:focus{
outline: none;
}
.srch:focus{
outline: none;
}
.content{
width: 1200px;
height: auto;
margin: auto;
color: #fff;
position: relative;
}
.content .par{
padding-left: 20px;
padding-bottom: 25px;
font-family: Arial;
letter-spacing: 1.2px;
line-height: 30px;
}
```

```
.content h1{
  font-family: 'Times New Roman';
  font-size: 50px;
  padding-left: 20px;
  margin-top: 9%;
  letter-spacing: 2px;
}
.content .cn{
  width: 160px;
  height: 40px;
  background: #ff7200;
  border: none;
  margin-bottom: 10px;
  margin-left: 20px;
  font-size: 18px;
  border-radius: 10px;
  cursor: pointer;
  transition: .4s ease;
}
.content .cn a{
  text-decoration: none;
  color: #000;
  transition: .3s ease;
}
.cn:hover{
  background-color: #fff;
}
.content span{
  color: #ff7200;
  font-size: 65px;
}
.form{
  width: 250px;
  height: 380px;
  background: linear-gradient(to top, rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
  position: absolute;
  top: -20px;
  left: 870px;
  transform: translate(0%,-5%);
  border-radius: 10px;
  padding: 25px;
}
.form h2{
  width: 220px;
```

```
font-family: sans-serif;
text-align: center;
color: #ff7200;
font-size: 22px;
background-color: #fff;
border-radius: 10px;
margin: 2px;
padding: 8px;
}
.form input{
width: 240px;
height: 35px;
background: transparent;
border-bottom: 1px solid #ff7200;
border-top: none;
border-right: none;
border-left: none;
color: #fff;
font-size: 15px;
letter-spacing: 1px;
margin-top: 30px;
font-family: sans-serif;
}
.form input:focus{
outline: none;
}
::placeholder{
color: #fff;
font-family: Arial;
}
.btnn{
width: 240px;
height: 40px;
background: #ff7200;
border: none;
margin-top: 30px;
font-size: 18px;
border-radius: 10px;
cursor: pointer;
color: #fff;
transition: 0.4s ease;
}
.btnn:hover{
background: #fff;
color: #ff7200;
```

```

}
.btnn a{
    text-decoration: none;
    color: #000;
    font-weight: bold;
}
.form .link{
    font-family: Arial, Helvetica, sans-serif;
    font-size: 17px;
    padding-top: 20px;
    text-align: center;
}
.form .link a{
    text-decoration: none;
    color: #ff7200;
}
.liw{
    padding-top: 15px;
    padding-bottom: 10px;
    text-align: center;
}
.icons a{
    text-decoration: none;
    color: #fff;
}
.icons ion-icon{
    color: #fff;
    font-size: 40px;
    padding-left: 60px;
    padding-top: 5px;
    transition: 0.3s ease;
}
.icons ion-icon:hover{
    color: #ff7200;
}
@media screen and (max-width:1200px) {
    /*Normal Screen*/

    .navbar{
        width: 100%;
        height: 100px;
    }
    ul{
        margin-left: 30px;
    }
}

```

```

ul li{
    margin-left: 60px;
}
ul li a{
    font-size: 1.6vw;
}
.search{
    margin-top: 3px;
    margin-left: 290px;
}
.srch{
    height: 40px;
    width: 190px;
    font-size: 14px;
}
.btn{
    height:40px;
    width: 80px;
}
.content{
    width: 100%;
}
.content h1, .content span{
    font-size: 4.5vw;
}
.content .par{
    width: 90%;
    font-size: 1.5vw;
}
.content .cn{
    width: 13%;
    height: 3.5vw;
    font-size: 1.8vw;
}
.content a{
    font-size: 1.6vw
}
}

@media screen and (max-width:1170px) {
    /*Login-box*/
    .main{
        padding-left: 20px;
        height: 180vh;
    }
}

```

```
.form{
  margin-left: -30px;
  width: 250px;
  height: 370px;
  background: linear-gradient(to top,
rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
  position: absolute;
  top: 420px;
  left: 50px;
  transform: translate(0%,-5%);
  border-radius: 10px;
  padding: 25px;
}
.form input{
  width: 240px;
  height: 35px;
  background: transparent;
  border-bottom: 1px solid #ff7200;
  border-top: none;
  border-right: none;
  border-left: none;
  color: #fff;
  font-size: 15px;
  letter-spacing: 1px;
  margin-top: 30px;
  font-family: sans-serif;
}
.btnn a{
  font-size: 16px;
}
.form .link a{
  font-size: 16px;
}
}
```

```
@media screen and (max-width: 830px){
  /*For tablet*/
```

```
.content{
  margin-top: 120px;
  width: 80%;
  margin-left: 40px;
}
.content h1, .content span{
  font-size: 6vw;
```



```
}
.content .par{
    width: 90%;
    font-size: 1.8vw;
}
.content .cn{
    width: 15%;
    height: 4.5vw;
    font-size: 2vw;
}
.content a{
    font-size: 2vw
}
.logo{
    margin-left: 240px;
    width: 100%;
    margin-top: 15px;
    font-size: 5vw;
}
.menu{
    width: 100%;
}
ul{
    margin-top: -5px;
    margin-left: 5px;
}
ul li{
    margin-left: 60px;
}
ul li a{
    font-size: 2vw;
}
.search{
    margin-top: -20px;
    margin-left: 60px;
}
.srch{
    height: 30px;
    width: 160px;
    font-size: 12px;
}
.btn{
    height: 30px;
    width: 70px;
}
```

```

.main{
  padding-left: 20px;
  height: 180vh;
}
.form{
  margin-left: -30px;
  width: 250px;
  height: 370px;
  background: linear-gradient(to top,
rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
  position: absolute;
  top: 430px;
  left: 50px;
  transform: translate(0%,-5%);
  border-radius: 10px;
  padding: 25px;
}
.form input{
  width: 240px;
  height: 35px;
  background: transparent;
  border-bottom: 1px solid #ff7200;
  border-top: none;
  border-right: none;
  border-left: none;
  color: #fff;
  font-size: 15px;
  letter-spacing: 1px;
  margin-top: 30px;
  font-family: sans-serif;
}
.btnn a{
  font-size: 16px;
}
.form .link a{
  font-size: 16px;
}
}

```

```

@media screen and (max-width: 600px){
  /*IPAD*/

```

```

  .content{
    margin-top: 80px;
    margin-left: 20px;

```

```

    }
    .search{
        margin-top: -40px;
        margin-left: 42px;
    }
    .logo{
        margin-left: 180px;
        font-size: 4vw;
    }
    ul{
        margin-top: -25px;
        margin-left: -5px;
    }
    ul li {
        margin-left: 50px;
    }
    ul li a{
        font-size: 2vw;
    }
}

@media screen and (max-width: 450px){
    /*mobile*/
    .logo{
        margin-left: 140px;
        font-size: 4vw;
    }
    ul{
        margin-top: -25px;
    }
    ul li {
        margin-left: 42px;
    }
    ul li a{
        font-size: 2vw;
    }
    .search{
        margin-top: -40px;
        margin-left: 38px;
    }
}

```

## **.REGISTER PAGE:**

HTML

[illegible]

```

        <span id="male">Male</span>
        &nbsp; &nbsp; &nbsp; &nbsp;
        <input type="radio" name="gender"
        id="female">
        &nbsp; &nbsp;
        <span id="female">Female</span>
        <br><br>
        <input type="submit" value="Submit"
        name="submit" id="submit"></center>
    </form>
</div>
</div>
</body>
</html>

```

CSS

```

*{
    margin: 0;
    padding: 0;
}

.main{
    width: 100%;
    background:url(crud.webp);
    background-position: center;
    background-size: cover;
    height: 100vh;
}

.navbar{
    width: 1200px;
    height: 75px;
    margin: auto;
}

.icon{
    width: 200px;
    float: left;
    height: 70px;
}

.logo{
    color: #ff7200;
    font-size: 35px;
    font-family: Arial;
    padding-left: 20px;
    float: left;
}

```

```
padding-top: 10px;
margin-top: 5px
}
.menu{
width: 400px;
float: left;
height: 70px;
}
ul{
float: left;
display: flex;
justify-content: center;
align-items: center;
}
ul li{
list-style: none;
margin-left: 62px;
margin-top: 27px;
font-size: 14px;
}
ul li a{
text-decoration: none;
color: #fff;
font-family: Arial;
font-weight: bold;
transition: 0.4s ease-in-out;
}
ul li a:hover{
color: #ff7200;
}
.search{
width: 330px;
float: left;
margin-left: 270px;
}
.srch{
font-family: 'Times New Roman';
width: 200px;
height: 40px;
background: transparent;
border: 1px solid #ff7200;
margin-top: 13px;
color: #fff;
border-right: none;
font-size: 16px;
```

```

float: left;
padding: 10px;
border-bottom-left-radius: 5px;
border-top-left-radius: 5px;
}
.btn{
width: 100px;
height: 40px;
background: #ff7200;
border: 2px solid #ff7200;
margin-top: 13px;
color: #fff;
font-size: 15px;
border-bottom-right-radius: 5px;
border-bottom-right-radius: 5px;
transition: 0.2s ease;
cursor: pointer;
}
.btn:hover{
color: white;
}
.btn:focus{
outline: none;
}
.srch:focus{
outline: none;
}
}
.content{
width: 1200px;
height: auto;
margin: auto;
color: #fff;
position: relative;
}
.content .par{
padding-left: 20px;
padding-bottom: 25px;
font-family: Arial;
letter-spacing: 1.2px;
line-height: 30px;
}
.content h1{
font-family: 'Times New Roman';
font-size: 50px;
padding-left: 20px;

```

```
margin-top: 9%;
letter-spacing: 2px;
}
.content .cn{
width: 160px;
height: 40px;
background: #ff7200;
border: none;
margin-bottom: 10px;
margin-left: 20px;
font-size: 18px;
border-radius: 10px;
cursor: pointer;
transition: .4s ease;
}
.content .cn a{
text-decoration: none;
color: #000;
transition: .3s ease;
}
.cn:hover{
background-color: #fff;
}
.content span{
color: #ff7200;
font-size: 65px
}
.form{
width: 250px;
height: 380px;
background: linear-gradient(to top, rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
position: absolute;
top: -20px;
left: 870px;
transform: translate(0%,-5%);
border-radius: 10px;
padding: 25px;
}
.form h2{
width: 220px;
font-family: sans-serif;
text-align: center;
color: #ff7200;
font-size: 22px;
```



```
background-color: #fff;
border-radius: 10px;
margin: 2px;
padding: 8px;
}
.form input{
width: 240px;
height: 35px;
background: transparent;
border-bottom: 1px solid #ff7200;
border-top: none;
border-right: none;
border-left: none;
color: #fff;
font-size: 15px;
letter-spacing: 1px;
margin-top: 30px;
font-family: sans-serif;
}
.form input:focus{
outline: none;
}
::placeholder{
color: #fff;
font-family: Arial;
}
.btnn{
width: 240px;
height: 40px;
background: #ff7200;
border: none;
margin-top: 30px;
font-size: 18px;
border-radius: 10px;
cursor: pointer;
color: #fff;
transition: 0.4s ease;
}
.btnn:hover{
background: #fff;
color: #ff7200;
}
.btnn a{
text-decoration: none;
color: #000;
```

```

        font-weight: bold;
    }
    .form .link{
        font-family: Arial, Helvetica, sans-serif;
        font-size: 17px;
        padding-top: 20px;
        text-align: center;
    }
    .form .link a{
        text-decoration: none;
        color: #ff7200;
    }
    .liw{
        padding-top: 15px;
        padding-bottom: 10px;
        text-align: center;
    }
    .icons a{
        text-decoration: none;
        color: #fff;
    }
    .icons ion-icon{
        color: #fff;
        font-size: 40px;
        padding-left: 60px;
        padding-top: 5px;
        transition: 0.3s ease;
    }
    .icons ion-icon:hover{
        color: #ff7200;
    }
    @media screen and (max-width:1200px) {
        /*Normal Screen*/

        .navbar{
            width: 100%;
            height: 100px;
        }
        ul{
            margin-left: 30px;
        }
        ul li{
            margin-left: 60px;
        }
        ul li a{

```

```

        font-size: 1.6vw;
    }
    .search{
        margin-top: 3px;
        margin-left: 290px;
    }
    .srch{
        height: 40px;
        width: 190px;
        font-size: 14px;
    }
    .btn{
        height:40px;
        width: 80px;
    }
    .content{
        width: 100%;
    }
    .content h1, .content span{
        font-size: 4.5vw;
    }
    .content .par{
        width: 90%;
        font-size: 1.5vw;
    }
    .content .cn{
        width: 13%;
        height: 3.5vw;
        font-size: 1.8vw;
    }
    .content a{
        font-size: 1.6vw
    }
}

@media screen and (max-width:1170px) {
    /*Login-box*/
    .main{
        padding-left: 20px;
        height: 180vh;
    }
    .form{
        margin-left: -30px;
        width: 250px;
        height: 370px;
    }
}

```

```

        background: linear-gradient(to top,
        rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
        position: absolute;
        top: 420px;
        left: 50px;
        transform: translate(0%,-5%);
        border-radius: 10px;
        padding: 25px;
    }
    .form input{
        width: 240px;
        height: 35px;
        background: transparent;
        border-bottom: 1px solid #ff7200;
        border-top: none;
        border-right: none;
        border-left: none;
        color: #fff;
        font-size: 15px;
        letter-spacing: 1px;
        margin-top: 30px;
        font-family: sans-serif;
    }
    .btnn a{
        font-size: 16px;
    }
    .form .link a{
        font-size: 16px;
    }
}

```

```

@media screen and (max-width: 830px){
    /*For tablet*/

```

```

    .content{
        margin-top: 120px;
        width: 80%;
        margin-left: 40px;
    }
    .content h1, .content span{
        font-size: 6vw;
    }
    .content .par{
        width: 90%;
        font-size: 1.8vw;
    }

```

```
}  
.content .cn{  
  width: 15%;  
  height: 4.5vw;  
  font-size: 2vw;  
}  
.content a{  
  font-size: 2vw  
}  
.logo{  
  margin-left: 240px;  
  width: 100%;  
  margin-top: 15px;  
  font-size: 5vw;  
}  
.menu{  
  width: 100%;  
}  
ul{  
  margin-top: -5px;  
  margin-left: 5px;  
}  
ul li{  
  margin-left: 60px;  
}  
ul li a{  
  font-size: 2vw;  
}  
.search{  
  margin-top: -20px;  
  margin-left: 60px;  
}  
.srch{  
  height: 30px;  
  width: 160px;  
  font-size: 12px;  
}  
.btn{  
  height: 30px;  
  width: 70px;  
}  
.main{  
  padding-left: 20px;  
  height: 180vh;  
}
```

```

.form{
  margin-left: -30px;
  width: 250px;
  height: 370px;
  background: linear-gradient(to top,
rgba(0,0,0,0.8)50%,rgba(0,0,0,0.8)50%);
  position: absolute;
  top: 430px;
  left: 50px;
  transform: translate(0%,-5%);
  border-radius: 10px;
  padding: 25px;
}
.form input{
  width: 240px;
  height: 35px;
  background: transparent;
  border-bottom: 1px solid #ff7200;
  border-top: none;
  border-right: none;
  border-left: none;
  color: #fff;
  font-size: 15px;
  letter-spacing: 1px;
  margin-top: 30px;
  font-family: sans-serif;
}
.btnn a{
  font-size: 16px;
}
.form .link a{
  font-size: 16px;
}
}

```

```

@media screen and (max-width: 600px){
  /*IPAD*/

```

```

  .content{
    margin-top: 80px;
    margin-left: 20px;
  }
  .search{
    margin-top: -40px;
    margin-left: 42px;

```

```

    }
    .logo{
        margin-left: 180px;
        font-size: 4vw;
    }
    ul{
        margin-top: -25px;
        margin-left: -5px;
    }
    ul li {
        margin-left: 50px;
    }
    ul li a{
        font-size: 2vw;
    }
}

@media screen and (max-width: 450px){
    /*mobile*/
    .logo{
        margin-left: 140px;
        font-size: 4vw;
    }
    ul{
        margin-top: -25px;
    }
    ul li {
        margin-left: 42px;
    }
    ul li a{
        font-size: 2vw;
    }
    .search{
        margin-top: -40px;
        margin-left: 38px;
    }
}

```

## PREDICTED PRICE:

```

<!DOCTYPE html>
<html>
<head>

```

```

<title>PRICE PREDICTION</title>
</head>
<body>
<p>&nbsp;</p>

<p>&nbsp;</p>
<style>
body {
  background-image: url(https://encrypted-
tbn0.gstatic.com/images?q=tbn:ANd9GcQdHJLQ-L0QqG6qGrs-
0WmAkdnQlo_9PsblfA&usqp=CAU);
  background-repeat: no-repeat;
  background-attachment: fixed;
  background-size: cover;
}
{
  min-height: 70%;
}
body,form {
padding: 0;
margin: 0;
outline: none;
font-family: Roboto, Arial, sans-serif;
font-size: 14px;
color: darkorange;
line-height: 22px;
}
.myDiv {
border: 5px outset #000000;
background-color: greenyellow;
text-align: center;
font-family: Roboto, Arial, sans-serif;
font-size: 14px;
color: red;
}
</style>

<h1 style="text-align:center"><span style="color:RED"><strong><span style="font-
family:Arial,Helvetica,sans-serif"><span style="font-size:36px">CRUDE OIL PRICE
PREDICTION&nbsp;</span></span></strong></span></h1>
<p>&nbsp;</p>
<p>&nbsp;</p>
<p>&nbsp;</p>
<form style="text-align:center">
  <label for="ENTER PRICE">ENTER PRICE:</label>
  <input type="date" id="PRICE" name="PRICE">

```



```
<input type="submit">
</form>

<p>&nbsp;</p>
<p>&nbsp;</p>

<div class="myDiv">
  <h2>PREDICTED PRICE:</h2>
  <p>{_____}</P><br>

  </p>
</div>

</body>
</html>
```

## 8. TESTING

### 8.1 Test case

Test case analysis

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fail	Pass
ML Model	4	0	0	4
Flask Application	4	0	0	4
IBM cloud	4	0	0	4
Exception Reporting	2	0	0	2
Final Report output	4	0	0	4

## **8.2 User Acceptance Testing**

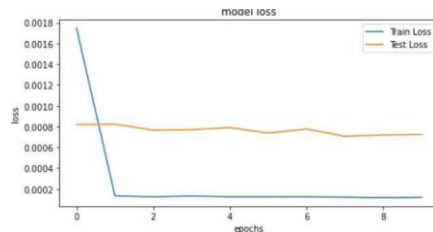
The purpose is to briefly explain the test coverage and open issues of the crude oil price prediction project at the time of the release to user acceptance testing

Defect Analysis:

The report shows the number of resolved and closed bugs at each severity level and how they were resolved

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	3	0	0	0	3
Duplicate	1	0	1	0	2
External	0	0	0	0	0
Fixed	4	0	1	1	6

S.No	Parameters	Values	Screenshot																		
1.	Model Summary		<div>Model: "sequential_1"</div> <table><thead><tr><th>Layer (type)</th><th>Output Shape</th><th>Param #</th></tr></thead><tbody><tr><td colspan="3">=====</td></tr><tr><td>lstm_3 (LSTM)</td><td>(None, 10, 50)</td><td>10400</td></tr><tr><td>lstm_4 (LSTM)</td><td>(None, 10, 50)</td><td>20200</td></tr><tr><td>lstm_5 (LSTM)</td><td>(None, 50)</td><td>20200</td></tr><tr><td>dense_1 (Dense)</td><td>(None, 1)</td><td>51</td></tr></tbody></table> <div>=====</div> <div>Total params: 50,851 Trainable params: 50,851 Non-trainable params: 0</div>	Layer (type)	Output Shape	Param #	=====			lstm_3 (LSTM)	(None, 10, 50)	10400	lstm_4 (LSTM)	(None, 10, 50)	20200	lstm_5 (LSTM)	(None, 50)	20200	dense_1 (Dense)	(None, 1)	51
Layer (type)	Output Shape	Param #																			
=====																					
lstm_3 (LSTM)	(None, 10, 50)	10400																			
lstm_4 (LSTM)	(None, 10, 50)	20200																			
lstm_5 (LSTM)	(None, 50)	20200																			
dense_1 (Dense)	(None, 1)	51																			

2.	Accuracy		<p> Train Mean Absolute Error: 1.0278217422906264  Train Root Mean Squared Error: 1.4285248639934083  Test Mean Absolute Error: 2.780526920817909  Test Root Mean Squared Error: 3.6348234466523737 </p> 
----	----------	--	---

Not Reproduced	0	0	0	0	0
Skipped	0	0	0	0	0
Won't fix	0	0	0	1	1
Totals	8	0	2	2	12

#### Test case analysis

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fail	Pass
ML Model	4	0	0	4
Flask Application	4	0	0	4
IBM Cloud	4	0	0	4
Exception Reporting	2	0	0	2
Final Report Output	4	0	0	4

## **9.RESULT**

### **9.1 Performance Metrics**

## **10.ADVANTAGES & DISADVANTAGES**

ADVANTAGES

- Prediction of crude oil price can help the importers to choose the right time to buy as they wait for the prices to fall down
- Prediction of crude oil prices can help the exporters to increase the demand
- It can even help in shifting the political powers
- can assist in minimizing the risks associated with volatility in oil prices

#### DISADVANTAGES

- The prediction results may lack accuracy
- Volatility in prices may be misleading

## **11.CONCLUSION**

LSTM network is better than other traditional neural networks for forecasting prices as it aims in using a back propagation model. Traditional neural networks such as CNN on the other hand predicts the next outgoing but doesn't necessarily save the previous data or connection which is based on feedforwarding, in the sense the previous data is not necessary to predict the future data. LSTM focuses on storing the previous data and prediction which is rather encouraging and more approximate. The outcomes derived are relatively encouraging. The results show that large lookups do not necessarily improve the accuracy of the predictions of crude oil prices. Hence it can be concluded, the model with a single LSTM model is definitely the most accurate

## **12.FUTURE SCOPE**

The project's future potential is enormous. The project can be implemented with the real-time functionalities that are necessary. Because it is quite versatile in terms of expansion, the project can be upgraded in the near future as and when the need arises. The complete prediction value can be increased in a much better, accurate, and error-free manner with the proposed approach. The project can be enhanced with real time data.

## **APPENDIX**

### **SOURCE CODE**

#### **MODEL:**

```
#DATA PREPROCESSING
```

```
#IMPORTING THE LIBRARIES
```

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pandas as pd\n",
      "import numpy
as np\n",
      "import matplotlib.pyplot as plt"
    ]
  }
]
```

#HANDLING MISSING VALUES

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    }
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      "cell_type": "code",
      "source": [
        "import
pandas as pd\n",
        "import numpy
as np\n",
        "import matplotlib.pyplot as plt"
```



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  "outputs": []
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  "metadata": {
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    }
  },
}
```

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"outputs": [
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    "data": {
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        "Closing Value  True\n",
        "dtype: bool"
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    "metadata": {},
    "execution_count": 3
  }
],
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  "metadata": {
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```

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  "outputId": "4001984b-54cc-4b54-f73b-926ff03ac00f"
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    "data": {
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        "Closing Value  7\n",
        "dtype: int64"
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    },
    "metadata": {},
    "execution_count": 4
  }
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  ],

```

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    "outputId": "47f79e73-6c7b-4d04-a0bd-09ae39d18bcf"
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      "data": {
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```

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"metadata": {},
"execution_count": 6
}
]
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    ],
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        "outputId": "2011717f-466c-4af2-973c-3980b6229a4d",
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        }
    },
    "execution_count": 7,
    "outputs": [
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```

```

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  "1    26.00\n",
  "2    26.53\n",
  "3    25.85\n",
  "4    25.87\n",
  "    ... \n",
  "8211   73.89\n",
  "8212   74.19\n",
  "8213   73.05\n",
  "8214   73.78\n",
  "8215   73.93\n",
  "Name: Closing Value, Length: 8216, dtype: float64"
]
},
"metadata": {},
"execution_count": 7
}
]
}
]
}

```

#FEATURE SCALING

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  "language_info": {
    "name": "python"
  }
},
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  {
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    "source": [
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pandas as pd\n",
      "import numpy
as np\n",
      "import matplotlib.pyplot as plt"
    ],
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    },
  },
]
```

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"outputs": []
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  "metadata": {
    "id": "dbaHUfMiJW8I"
  },
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  "outputs": []
},
{
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    "data.isnull().any()"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    },
    "id": "RAUyZB0-Jp0t",
    "outputId": "9f1ec869-c35f-4764-8989-0d5d82646e18"
  },
}
```



```
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        "Closing Value   True\n",
        "dtype: bool"
      ]
    },
    "metadata": {},
    "execution_count": 3
  }
],
{
  "cell_type": "code",
  "source": [
    "data.isnull().sum()"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
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    "output_type": "execute_result",
    "data": {
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        "Closing Value  7\n",
        "dtype: int64"
      ]
    },
    "metadata": {},
    "execution_count": 4
  }
],
{
  "cell_type": "code",
  "source": [
    "data.dropna(axis=0,inplace=True)"
  ],
  "metadata": {
    "id": "g3FWuWVPJwms"
  },
}
```

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"execution_count": 5,
"outputs": []
},
{
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}
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]
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  },
  "execution_count": 7,
  "outputs": [
    {
      "output_type": "execute_result",
      "data": {
        "text/plain": [
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          "1    26.00\n",
```

```

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"3    25.85\n",
"4    25.87\n",
"    ... \n",
"8211  73.89\n",
"8212  74.19\n",
"8213  73.05\n",
"8214  73.78\n",
"8215  73.93\n",
"Name: Closing Value, Length: 8216, dtype: float64"
]
},
"metadata": {},
"execution_count": 7
}
]
},
{
"cell_type": "code",
"source": [
"from sklearn.preprocessing import MinMaxScaler\n",
"scaler=MinMaxScaler(feature_range=(0,1))\n",
"data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))"
],
"metadata": {
"id": "5m-DUFI9J_WN"

```

```
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"outputs": []
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  ],
  "metadata": {
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    "outputId": "6e2bdf01-77aa-4e93-f117-6ab963e21c0a",
    "colab": {
      "base_uri": "https://localhost:8080/"
    }
  },
  "execution_count": 9,
  "outputs": [
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      "output_type": "execute_result",
      "data": {
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            "      [0.11661484],\n",
            "      [0.12053902],\n",
            "      ..., \n",
```

```

        "    [0.46497853],\n",
        "    [0.47038353],\n",
        "    [0.47149415]])"
    ]
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"metadata": {},
"execution_count": 9
}
]
}
]
}

```

## # DATA VISUALIZATION

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pandas as pd\n",
      "import numpy
as np\n",
      "import matplotlib.pyplot as plt"
    ],
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    },
    "execution_count": 1,
    "outputs": []
  },
  {
    "cell_type": "code",
    "source": [
      "data=pd.read_excel(\"/content/Crude Oil Prices Daily.xlsx\")"
    ],
    "metadata": {
      "id": "dbaHUfMiJW8I"
```



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  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/"
    },
    "id": "RAUyZB0-Jp0t",
    "outputId": "9f1ec869-c35f-4764-8989-0d5d82646e18"
  },
  "execution_count": 3,
  "outputs": [
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      "output_type": "execute_result",
      "data": {
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          "Closing Value  True\n",
          "dtype: bool"
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      }
    ]
  ]
}
```

```
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    "execution_count": 3
  }
]
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    "data.isnull().sum()"
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    "outputId": "4001984b-54cc-4b54-f73b-926ff03ac00f"
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  "execution_count": 4,
  "outputs": [
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```

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}
]
},
{
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  ],
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  },
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    "data.isnull().sum()"
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    "colab": {
```

```
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  "outputId": "47f79e73-6c7b-4d04-a0bd-09ae39d18bcf"
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        "Closing Value  0\n",
        "dtype: int64"
      ]
    },
    "metadata": {},
    "execution_count": 6
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],
{
  "cell_type": "code",
  "source": [
    "data_oil=data.reset_index()['Closing Value']\n",
    "data_oil"
```

```

],
"metadata": {
  "colab": {
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  },
  "id": "IoxUNwrvJ2b6",
  "outputId": "2011717f-466c-4af2-973c-3980b6229a4d"
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"execution_count": 7,
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    "output_type": "execute_result",
    "data": {
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        "1    26.00\n",
        "2    26.53\n",
        "3    25.85\n",
        "4    25.87\n",
        "    ... \n",
        "8211  73.89\n",
        "8212  74.19\n",
        "8213  73.05\n",
        "8214  73.78\n",
        "8215  73.93\n",
        "Name: Closing Value, Length: 8216, dtype: float64"
      ]
    }
  ]

```

```

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    "metadata": {},
    "execution_count": 7
  }
]
},
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    "scaler=MinMaxScaler(feature_range=(0,1))\n",
    "data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))"
  ],
  "metadata": {
    "id": "5m-DUFI9J_WN"
  },
  "execution_count": 8,
  "outputs": []
},
{
  "cell_type": "code",
  "source": [
    "data_oil"
  ],
  "metadata": {
    "colab": {

```

```
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},
"id": "fyMwOo1jKLqL",
"outputId": "6e2bdf01-77aa-4e93-f117-6ab963e21c0a"
},
"execution_count": 9,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "array([[0.11335703],\n",
        "       [0.11661484],\n",
        "       [0.12053902],\n",
        "       ..., \n",
        "       [0.46497853],\n",
        "       [0.47038353],\n",
        "       [0.47149415]])"
      ]
    },
    "metadata": {},
    "execution_count": 9
  }
],
},
{
```

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  "outputId": "0640d4b9-abe5-4e76-a910-871af84bc19c",
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    "output_type": "execute_result",
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      ]
    },
    "metadata": {},
    "execution_count": 10
  },
  {
    "output_type": "display_data",
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  "image/png":
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## #CREATING A DATA SET WITH SLIDING WINDOWS

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      "display_name": "Python 3"
    },
    "language_info": {
      "name": "python"
    }
  },
  "cells": [
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      "cell_type": "code",
      "source": [
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```
"import
pandas as pd\n",
"import numpy
as np\n",
    "import matplotlib.pyplot as plt"
],
"metadata": {
    "id": "GiRQ27X4JRcH"
},
"execution_count": 1,
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},
{
    "cell_type": "code",
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        "data=pd.read_excel(\"/content/Crude Oil Prices Daily.xlsx\")"
    ],
    "metadata": {
        "id": "dbaHUfMiJW8I"
    },
    "execution_count": 2,
    "outputs": []
},
{
    "cell_type": "code",
    "source": [
```

```
"data.isnull().any()"
],
"metadata": {
  "colab": {
    "base_uri": "https://localhost:8080/"
  },
  "id": "RAUyZB0-Jp0t",
  "outputId": "9f1ec869-c35f-4764-8989-0d5d82646e18"
},
"execution_count": 3,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "Date          False\n",
        "Closing Value  True\n",
        "dtype: bool"
      ]
    },
    "metadata": {},
    "execution_count": 3
  }
],
{
```

```
"cell_type": "code",
"source": [
  "data.isnull().sum()"
],
"metadata": {
  "colab": {
    "base_uri": "https://localhost:8080/"
  },
  "id": "yorF39lCJsV",
  "outputId": "4001984b-54cc-4b54-f73b-926ff03ac00f"
},
"execution_count": 4,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "Date          0\n",
        "Closing Value  7\n",
        "dtype: int64"
      ]
    },
    "metadata": {},
    "execution_count": 4
  }
]
```

```
},  
{  
  "cell_type": "code",  
  "source": [  
    "data.dropna(axis=0,inplace=True)"  
  ],  
  "metadata": {  
    "id": "g3FWuWVPJwms"  
  },  
  "execution_count": 5,  
  "outputs": []  
},  
{  
  "cell_type": "code",  
  "source": [  
    "data.isnull().sum()"  
  ],  
  "metadata": {  
    "colab": {  
      "base_uri": "https://localhost:8080/"  
    },  
    "id": "vIHbyOs4JzIf",  
    "outputId": "47f79e73-6c7b-4d04-a0bd-09ae39d18bcf"  
  },  
  "execution_count": 6,  
  "outputs": [
```

```

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  "data": {
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      "Closing Value  0\n",
      "dtype: int64"
    ]
  },
  "metadata": {},
  "execution_count": 6
}
],
{
  "cell_type": "code",
  "source": [
    "data_oil=data.reset_index()['Closing Value']\n",
    "data_oil"
  ],
  "metadata": {
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      "base_uri": "https://localhost:8080/"
    },
    "id": "IoxUNwrvJ2b6",
    "outputId": "2011717f-466c-4af2-973c-3980b6229a4d"
  }
}

```

```

},
"execution_count": 7,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "0    25.56\n",
        "1    26.00\n",
        "2    26.53\n",
        "3    25.85\n",
        "4    25.87\n",
        "    ... \n",
        "8211  73.89\n",
        "8212  74.19\n",
        "8213  73.05\n",
        "8214  73.78\n",
        "8215  73.93\n",
        "Name: Closing Value, Length: 8216, dtype: float64"      ]
      },
    },
    "metadata": {},
    "execution_count": 7
  }
],
},
{

```

```

"cell_type": "code",
"source": [
    "from sklearn.preprocessing import MinMaxScaler\n",
    "scaler=MinMaxScaler(feature_range=(0,1))\n",
    "data_oil=scaler.fit_transform(np.array(data_oil).reshape(-1,1))"
],
"metadata": {
    "id": "5m-DUFI9J_WN"
},
"execution_count": 8,
"outputs": []
},
{
    "cell_type": "code",
    "source": [
        "data_oil"
    ],
    "metadata": {
        "colab": {
            "base_uri": "https://localhost:8080/"
        },
        "id": "fyMwOo1jKLqL",
        "outputId": "6e2bdf01-77aa-4e93-f117-6ab963e21c0a"
    },
    "execution_count": 9,
    "outputs": [

```



```

{
  "output_type": "execute_result",
  "data": {
    "text/plain": [
      "array([[0.11335703],\n",
      "      [0.11661484],\n",
      "      [0.12053902],\n",
      "      ..., \n",
      "      [0.46497853],\n",
      "      [0.47038353],\n",
      "      [0.47149415]])"
    ]
  },
  "metadata": {},
  "execution_count": 9
}

],
{
  "cell_type": "code",
  "source": [
    "plt.plot(data_oil)"
  ],
  "metadata": {
    "colab": {
      "base_uri": "https://localhost:8080/",

```

```
"height": 282
},
"id": "GdNJartuKUfk",
"outputId": "0640d4b9-abe5-4e76-a910-871af84bc19c"
},
"execution_count": 10,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "[<matplotlib.lines.Line2D at 0x7f5a0c7b5850>]"
      ]
    },
    "metadata": {},
    "execution_count": 10
  },
  {
    "output_type": "display_data",
    "data": {
      "text/plain": [
        "<Figure size 432x288 with 1 Axes>"
      ],
      "image/png":
    },
    "metadata": {
```

```

        "needs_background": "light"
    }
}
]
},
{
    "cell_type": "code",
    "source": [
        "training_size=int(len(data_oil)*0.65)\n",
        "test_size=len(data_oil)-training_size\n",

"train_data,test_data=data_oil[0:training_size:],data_oil[training_size:len
(data_oil),:1]"
    ],
    "metadata": {
        "id": "S9THgFu3KaKH"
    },
    "execution_count": 11,
    "outputs": []
},
{
    "cell_type": "code",
    "source": [
        "training_size,test_size"
    ],
    "metadata": {

```

```
"colab": {
  "base_uri": "https://localhost:8080/"
},
"id": "mbvhl2HCKmSG",
"outputId": "5142d0f7-21d1-46bb-e69e-bd71c69ef729"
},
"execution_count": 12,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "(5340, 2876)"
      ]
    },
    "metadata": {},
    "execution_count": 12
  }
],
{
  "cell_type": "code",
  "source": [
    "train_data.shape"
  ],
  "metadata": {
```

```
"colab": {
  "base_uri": "https://localhost:8080/"
},
"id": "M14MiYVzKrJg",
"outputId": "b23a9806-135a-424d-f65b-6029116ad975"
},
"execution_count": 13,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "(5340, 1)"
      ]
    },
    "metadata": {},
    "execution_count": 13
  }
],
{
  "cell_type": "code",
  "source": [
    "def create_dataset(dataset,time_step=1):\n",
    "    dataX,dataY=[],[]\n",
    "    for i in range(len(dataset)-time_step-1):\n",
```

```

    "    a=dataset[i:(i+time_step),0]\n",
    "    dataX.append(a)\n",
    "    dataY.append(dataset[i+time_step,0])\n",
    "    return np.array(dataX),np.array(dataY)"
],
"metadata": {
    "id": "vrH4kcM5K9S8"
},
"execution_count": 14,
"outputs": []
},
{
    "cell_type": "code",
    "source": [
        "time_step=10\n",
        "x_train,y_train=create_dataset(train_data,time_step)\n",
        "x_test,y_test=create_dataset(test_data,time_step)"
    ],
    "metadata": {
        "id": "dv2OIjslLA3v"
    },
    "execution_count": 15,
    "outputs": []
},
{
    "cell_type": "code",

```

```
"source": [  
  "print(x_train.shape),print(y_train.shape)"  
],  
"metadata": {  
  "colab": {  
    "base_uri": "https://localhost:8080/"  
  },  
  "id": "ApfZGkVhLD2p",  
  "outputId": "979643ec-00c5-4f15-93f0-8529e83335a0"  
},  
"execution_count": 16,  
"outputs": [  
  {  
    "output_type": "stream",  
    "name": "stdout",  
    "text": [  
      "(5329, 10)\n",  
      "(5329,)\n"  
    ]  
  },  
  {  
    "output_type": "execute_result",  
    "data": {  
      "text/plain": [  
        "(None, None)"  
      ]  
    }
```

```
    },  
    "metadata": {},  
    "execution_count": 16  
  }  
]  
},  
{  
  "cell_type": "code",  
  "source": [  
    "print(x_test.shape),print(y_test.shape)"  
  ],  
  "metadata": {  
    "colab": {  
      "base_uri": "https://localhost:8080/"  
    },  
    "id": "iLQiXr21LE35",  
    "outputId": "3dca6c25-95ed-4b85-bb1a-d2e218179c5d"  
  },  
  "execution_count": 17,  
  "outputs": [  
    {  
      "output_type": "stream",  
      "name": "stdout",  
      "text": [  
        "(2865, 10)\n",  
        "(2865,)\n"
```



```
]
},
{
  "output_type": "execute_result",
  "data": {
    "text/plain": [
      "(None, None)"
    ]
  },
  "metadata": {},
  "execution_count": 17
}
],
{
  "cell_type": "code",
  "source": [
    "x_train"
  ],
  "metadata": {
    "id": "w95tBQ4gLJWj",
    "outputId": "f1567ef1-fb46-4909-857b-c3c34c6f1dd1",
    "colab": {
      "base_uri": "https://localhost:8080/"
    }
  },
}
```

```
"execution_count": 18,
"outputs": [
  {
    "output_type": "execute_result",
    "data": {
      "text/plain": [
        "array([[0.11335703, 0.11661484, 0.12053902, ...,
0.10980305, 0.1089886 ,\n",          "      0.11054346],\n",
        "      [0.11661484, 0.12053902, 0.11550422, ..., 0.1089886 ,
0.11054346,\n",
        "      0.10165852],\n",
        "      [0.12053902, 0.11550422, 0.1156523 , ..., 0.11054346,
0.10165852,\n",
        "      0.09906708],\n",
        "      ..., \n",
        "      [0.36731823, 0.35176958, 0.36080261, ..., 0.36391234,
0.37042796,\n",
        "      0.37042796],\n",
        "      [0.35176958, 0.36080261, 0.35354657, ..., 0.37042796,
0.37042796,\n",
        "      0.37879461],\n",
        "      [0.36080261, 0.35354657, 0.35295424, ..., 0.37042796,
0.37879461,\n",
        "      0.37916482]])"
      ]
    },
    "metadata": {},
```

```

        "execution_count": 18
    }
]
},
{
    "cell_type": "code",
    "source": [
        "x_train=x_train.reshape(x_train.shape[0],x_train.shape[1],1)\n",
"x_test=x_test.reshape(x_test.shape[0],x_test.shape[1],1)"
    ],
    "metadata": {
        "id": "59VQDX1LLL3D"
    },
    "execution_count": 19,
    "outputs": []
}
]
}

```

## **# MODEL BUILDING**

```

# Importing the
model building
libraries from
tensorflow.keras.m
odels import
Sequential from
tensorflow.keras.la
yers import Dense

```

```

from
tensorflow.kera
s.layers import
LSTM      #
Initializing the
model
model=Sequen
tial()

```

### **# 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))

```

### **# Adding Output Layers**

```

model.add(Dense(1))

```

### **# Configure The Learning Process**

```

model.compile(loss='mean_squared_err
or',optimizer='ad am')

```

### **# Train The Model**

```

model.fit(X_train,y_train,validation_data=(X_test,y_test),epoch
s=10,batch_size=64,verbose=1)

```

### **# Model Evaluation**

```

train_predict=model.predict(X
_train)
test_predict=model.predict(X_t
est)

```

```

train_predict=scalar.inverse_transform(train_predict)
test_predict=scalar.inverse_transform(test_predict)
import math
from sklearn.metrics import mean_squared_error

math.sqrt(mean_squared_error(y_train,train_predict))

```

### # Save The Model

```

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

```

### # Test The Model

```

look_back= 10
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(scalar.inverse_transform(data_oil))
plt.plot(trainPredictPlot)
plt.plot(testPredictPlot)
plt.show()
len(test_data)
x_input=test_data[2866:].reshape(1,-1)
x_input.shape
ape
temp_input
t=list(x_input)

```

```

temp_inpu
t=temp_in
put[0].toli
st()
temp_inpu
t
lst_output
=
[]
n
-
s
t
e
p
s
=
1
0
i
=
0
w
h
i
l
e
(
i
<
1
0
)
:
    if(len(temp_input)>10):

```

```

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

```

1 else:

```

x_input =
x_input.reshape((1,
n_steps,1)) yhat =
model.predict(x_inpu
ut, verbose=0)
print(yhat[0])
temp_input.extend(
yhat[0].tolist())
print(len(temp_inpu
t))
lst_output.extend(y
hat.tolist())
i=i+1

print (lst_output)
day_new=np.arange(1,
11)
day_pred=np.arange(1
1,2
1) len(data_oil)
plt.plot(day_new,scalar.inverse_transform(data_oil[
8206:]
)) plt.title("Review of prediction")
plt.plot(day_pred,scalar.inverse_tr
ansform(lst_output)) plt.show()
df3=data_oil.tolist()

df3.extend(lst_outpu
t)
plt.title("Past data nad next 10
days output prediction")
plt.plot(df3[8100:])
df3=scalar.inverse_transform(df3)
.tolist()
plt.title("Past data nad next 10 days output prediction after reversing the
scaled values")

```



## **Index.html:**

```
<!DOCTYPE html>
<html lang="en">
  <head>
    <meta charset="UTF-8">
    <title>Crudeoil price prediction</title>
    <!--<link rel="stylesheet" href="in.css">
    <link
href="https://fonts.googleapis.com/css2?family=Poppins:wght
@300;400;500;600;
700&display=swap"rel="stylesheet">
    <link rel="stylesheet"
href="https://www.w3schools.com/w3css/4/w3.css">
<link rel="stylesheet"
href="https://cdnjs.cloudflare.com/ajax/libs/font-
awesome/4.7.0/css/fontawesome.min.c ss">-->
    <style
    >
    ul
    {
    li
    st
    -
    st
    yl
    e-
    ty
    p
    e:
    n
    o
```

n  
e;  
m  
ar  
gi  
n:  
0;  
p  
a  
d  
di  
n  
g:  
0;  
o  
v  
er  
fl  
o  
w  
:  
hi  
d  
d  
e  
n;  
b  
o  
r  
d  
er  
:  
1  
p  
x

s  
ol  
id  
#  
e  
7  
e  
7  
e  
7;  
b  
a  
c  
k  
g  
r  
o  
u  
n  
d  
c  
ol  
o  
r:  
#  
0  
5  
7  
5  
1  
4;  
}

```
i
{
  float: left;
}
```

```
li a {
  display:
  inlineblock;
  color: rgb(78,
  3, 3); text-
  align: center;
  padding: 14px
  16px;
  text-decoration
  : none;
  background-
  color:rgb(18,
  116, 5);
}
```

```
l
i
a
:
h
o
v
e
r
{
```

```

border:1px
solid;
background-
color:
lightseagreen; }
</style>
</head>
<body>
  <nav class="navbar navbar-inverse">
    <div class="container-fluid">
      <ul>
        <li class="parts"><a href="#">Home</a></li>
        <li class="parts"><a
href="predict.html">predict</a></li> </ul>
      </div>
    </nav>

    <h1 >Crudeoil price prediction</h1>

    <
      s
      t
      y
      l
      e
      >
      b
      o
      d
      y
      {
        background-image:
        url('static/css/image.j
        peg');
        backgroundrepeat:
        no-repeat;

```

```
background-  
attachment:  
fixed; background-size: 100%  
100%;  
  
}  
</style>
```

```
<h3 style="font-family:system-ui;">
```

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</h3  
>  
</body>  
</html>
```

### **Predict.html:**

```
<html>  
<head>  
<link rel="stylesheet" href="static/css/style.css">  
<  
s  
t  
y  
l
```

```
e
>
b
o
d
y
{
    background-image:
    url('static/css/image3.
    jpg');
    backgroundrepeat:
    no-repeat;
    background-
    attachment:
    fixed; background-size: 100%
    100%;

}
```

```
</head>
```

```
<script>
```

```
    document.getElementById("demo
    ").innerHTML =
document.getElementB
    yId("ten"); </script>
```

```
<body>
```

```
<form action="/method" method="POST" enctype =
"multipart/form-data">
```

```
<div class="container">
```

```
    <!--<div class="brand-logo"></div>-->
```

```
    <div class="brand-title">predict the oil price</div>
```

```
    <div class="inputs">
```

```
        <label>Enter Price</label>
```

```
        <input type="text" placeholder="Enter ten days price"
        id="ten" name="val"/> <button
```

```
        type="submit">Predict</b>
    </div>
    <div>
        <input type="text" value="0" />
    </div>
    <div>
        <input type="button" value="Predict" />
    </div>
    <div>
        <div>
            The next
            day price is :
            {{ prediction }}
        </div>
    </div>
</div>
</form>
</body>
</html>
```

### **App.py:**

```
from flask import Flask,
render_template, request, redirect
import numpy as np
# from tensorflow.k
```



```

from keras.saving.save import load_model
app = Flask(__name__, template_folder='template')

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

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

```

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

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

```
lst_output =
```

```
[  
]
```

```
n
```

```
-
```

```
s
```

```
t
```

```
e
```

```
p
```

```
s
```

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=
```

```
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```

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0
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0
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```

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l
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```
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```

```
(
```

```
i
```

```
<
```

```
1
```

```
0
```

```
)
```

```
:
```

```

if
    (len(temp_
input) >
10):
    x_input =
    np.array(te
mp_input[
1:])
    x_input =
    x_input.res
hape(1, -
1) x_input =
    x_input.reshape((
1, n_steps, 1))
    yhat =
    model.predict(x_i
nput, verbose=0)
    temp_input.exten
d(yhat[0].tolist())
    temp_input =
    temp_input[1:]
    lst_output.extend
(yhat.tolist())
    i = i + 1

e
l
s
e
:
    x_input =
    x_input.reshape((1,
n_steps, 1)) yhat =
    model.predict(x_inpu
t, verbose=0)

```

```

        temp_input.extend(yhat[0].tolist())
        lst_output.extend(yhat.tolist())
        i = i + 1
        val =
        lst_output[9]
    return render_template('predict.html', prediction=val)
if request.method == "GET":
    return render_template('predict.html')

if __name__ == "__main__":
    model =
    load_model(r'crudeoilpr
    ediction.h5')
    app.run(debug=True)

```

### **#cloud deployment code in ml model**

```

!pip install ibm_watson_machine_learning

from
ibm_watson_machine_learning
import APIClient
wml_credentials = {
    "url" : "https://us-south.ml.cloud.ibm.com",
    "apikey" :
    "cRkqykhsnLO1Ogs_xoYjgLkNTtTS1Qxyi0Mn1GSlQ1P5"
} client =
APIClient(wml
_credentials)

#for creating a
deployment
phase def
guid_from_spac

```

```

e_name(client,
space_name):
space=client.space
ces.get_details()
    #print(space) return(next(item
for item in space['resources'] if
item['entity']['name'] ==
space_name)['metadata']['id'])
space_uid =
guid_from_space_name(client,
'models') print("Space UID =
"+space_uid)
client.set.default_space(space_uid)
client.software_specifications.list
() software_spec_uid=
client.software_specifications.get_uid_by_name("tensorflow
rt22.1py3.9") software_spec_uid

```

**GitHub & Project Demo Link:**  
<https://github.com/IBM-EPBL/IBM-Project-47745-1660801886>