

CRUDE OIL PRICE PREDICTION

1. INTRODUCTION

Crude oil is the most important resources in today's world. Cost has the direct effect on the global habitat, our economy, political and other activities. Here the rise in price of oil will profit for the producers. The evaporation nature of crude oil, and its price prediction become extremely difficult tasks. We are here proposing the innovative method of predicting the crude oil price using the artificial neural network (ANN). These proposed system helps us to decide the proper crude oil price at correct time analysis.

2. OBJECTIVES:

BY THE END OF PROJECT WE WILL BE ABLE TO:

- Using the flask frame work we will be able to build an application.
- Understanding of time series data.
- Techniques and concepts of time series forecasting.
- Splits the data and analysis.

3. PROJECT FLOW:

TO COMPLETE CERTAIN ACTIVITIES AND TO ACCOMPLISH THE TASKS LISTED BELOW:

- Collection of data
- Dataset creation
- Preprocessing of data
- Libraries imports
- Dataset imports
- Data analyze
- Missing data
- Featuring the data scaling
- Visualizaton of the dataset and analyzing
- Splitting the data into test and train
- Sliding window of data set
- Model building
- Model building libraries importation

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- Initialize the model
- Adding LSTM and outline layers
- Learning model and training the process
- Evaluation of model
- Save the model
- Test the model
- Building application
- Create the HTML file
- Build the python code

4. DATA COLLECTION:

We have downloaded the dataset from kaggle:

<https://www.kaggle.com/rockbottom73/crude-oil-prices>

5. DATA PREPROCESSING:

Data pre processing includes the following tasks:

- Importing the libraries:

The required libraries to import the python scripts are,

5.1 NUMPY:

Its the open source numerical python script. It contains the multi dimensional array and matrix data structure also perform the mathematical operation on array such as trigonometric, statistical and algebra.

5.2 PANDAS:

One of the top python programming languages and it is fast , flexible and easy to use the open source data analysis.

5.3 MATPLOTLIP:

Creating static, animated and interactive visualizing of python

- Importing the dataset:
 - You might have your data in csv files, xlsx files
 - Loading the excel sheet into pandas
 - Locate the dictionary into the excel sheet to keep the data set more efficient .

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- If the data set is in some other location, we can find out with the help of pandas.
- Loading the pandas into same dictionary so we can read directly and easily.

5.4 HANDLING MISSING DATA:

- After loading the dataset check the rows and columns for their null values with complete information.
- If there are any null values, following can be done:
- Using data imputation the data is imputed in sklearn.
- Filling the NaN values with help of median, mode and mean using fillna() method.
- Delete the records.
- Now we can see the null values in the closing value column and also check how many numbers of null values in the column using sum() function.
- Drop the null from the column.
- Axis=0 drop the row.
- Data frame has to change permanent indicators, 'in place=True'.
- Reset_index consider the closing value column to reset the index of data frame list of integer from 0 to length of data.

5.5 FEATURE SCALING:

Feature scaling to normalize the independent variables to scale the crude oil prices between (0,1) to avoid the computation. Some of the common methods are standardization and Normalization.

5.6 STANDARDIZATION:

The process of developing and implementing the technical products and stability of the products.

5.7 NORMALIZATION:

The process of organizing the data base or data frame to reduce the redundancy and improve the integrity of the data, improve and simplifies the data design.

5.8 DATA VISUALIZATION:

- In data visualization the data set were presented in graphical pattern, it helps to detect the pattern, levels, correlation and trends in recent days which has not been detected in text format.

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- To understand the data algorithm and relationship even the well performed machine can also perform poorly which wasn't visualize properly and understood. We need libraries like Matplotlib and sea born to visualize the data set.

Matplotlib is the 2Dpython libraries which helps to plots, bar charts etc.,

Using the matplotlib and sea born libraries are:

Basic properties of plotting the graphs are:

- ☆ **X label:** set the x- axis.
- ☆ **Y label:**set the y-axis.
- ☆ **Title:** set the title for the axis.
- ☆ **Legend:**place the legend for the axes.

There are different types of analysis ;

UNIVARIATE:Using the single features to analyses the properties.

BIVARIATE: Compare the data between the 2 features.

MULTIVARIATE: Comparing more than 2 variables.

5.9 SPLITTING DATA INTO TRAIN AND TEST:

On working on model have to test and train the model and have to set the data frame . After training we have to test the some dataset.Train and test the dataset in the different set of data frames in the earlier ways. During the development phase there will be always the lost of dataset. So that the dataset have to split into two different set one is to training and other one is for testing.

For the time series data, and the data sequence are important. The size of train and test the data after splitting.

6 CREATING A DATASET WITH SLIDING WINDOW:

A special type of data set or structure needed to cover n-type stamp.LSTM will predict the n+1 price.The number of data stamp is set of 10.

The function has two arguments dataset which is Numpy want to convert the dataset and the time_step will predict the previous time set of the input variables, and

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defaulted to 1.

The default will create the dataset with X which will predict the price of crude oil at the time t , and Y will predict the crude oil at the next time $(n+t)$.

For LSTM model it requires the train and test of X in three dimensional array for building the model.

7 MODEL BUILDING:

Model buildings are includes,

1. Model building blocks importation.
2. Intialise the model
3. Add the LSTM layers
4. Add output layer
5. Learning process.
6. Training the model
7. Model evaluation.
8. Save the model
9. Test the model.

7.1 IMPORT THE MODEL BULIDING LIBRARIES:

Importing the necessary model libraries.

7.2 INTIALISE THE MODEL:

Linear stack of layer is the sequential model. Create the sequential order by passing the list of layer. Model imports the sequential order in keras.

7.3 ADDING THE LSTM LAYER:

Units is the number of LSTM neurons layer. Fifty neurons will give the model high dimensionality, enough to capture the upwards and downward trends.

Return_sequences is True as we need to add another LSTM layer after the current one. input_shape corresponds to the number of time stamps and the number of indicators.

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7.4 ADDING OUTPUT LAYER:

Dense layer is a deeply connected neural network layer. It is the most common and frequently used layer.

The output dimension is one since we are predicting one price each time.

Understanding the model is a crucial phase to use it for training properly and prediction purposes.

Keras provides a simple method, summary to get the full information about the model and its layers.

7.5 CONFIGURE THE LEARNING PROCESS:

Compilation is the final step in creating a model. Once the compilation is done, we can move on to the training phase

The loss function is used to find error or deviation in the learning process. Keras requires loss function during the model compilation process.

Optimization is an important process that optimizes the input weights by comparing the prediction and the loss function. Here we are using adam optimizer

Metrics are used to evaluate the performance of your model. It is similar to loss function, but not used in the training process

7.6 TRAIN THE MODEL:

RNN weights are updated every 64 stock prices with a batch size of 64. Try more batches and epochs if the loss of the model is not converging.

Arguments:

Epochs: an integer and number of epochs we want to train our model for.

An inputs and targets list

A generator

An inputs, targets, and sample_weights list which can be used to evaluate

The loss and metrics for any model after any epoch has ended.

8 MODEL EVALUATION:

we need to check to see how well our model is performing on the test data.

Regression Evaluation Metrics

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8.1 Mean Squared Error (MSE):

MSE or Mean Squared Error is one of the most preferred metrics for regression problems.

It is simply the average of the squared difference between the target value and the value predicted by the regression model.

As it squares the differences, it penalizes even a small error which leads to over-estimation of how bad the model is.

It is preferred more than other metrics because it is differentiable and hence can be optimized better.

8.2 RMSE:Root Mean Square Error:

RMSE is the square root of the averaged squared difference between the target value and the value predicted by the model.

It is preferred more in some cases because the errors are first squared before averaging which poses a high penalty on large errors.

This implies that RMSE is useful when large errors are undesired.

9 SAVE THE MODEL:

An H5 file is a data file saved in the Hierarchical Data Format (HDF). It contains multidimensional arrays of scientific data.

9.1 TEST THE MODEL:

Shift the predictions so that they align on the x-axis with the original dataset.

Once prepared, the data is plotted, showing the original dataset in blue, the predictions for the training dataset in green, the predictions on the unseen test dataset

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in orange.

10 APPLICATION BUILDING:

A UI is provided for the users where he has to enter the values for predictions.

The entered values are given to the saved model and prediction is showcased on the UI.

10.1 CREATE AN HTML FILE:

We use HTML to create the front end part of the web page.

Here, we created 2 HTML pages- index.html, web.html.

index.html displays the home page.

web.html accepts the values from the input and displays the prediction.

10.2 INDEX

```
<!DOCTYPE html>
<html lang="en" >
<head>
  <meta charset="UTF-8">
  <title>Sign Up Signin Form Template Example</title>
  <link rel="stylesheet" href="./style.css">

</head>
<body>
<!-- partial:index.partial.html -->
<html lang="en">
<head>
  <!-- Latest compiled and minified CSS -->
  <link
                                                                    rel="stylesheet"
href="https://maxcdn.bootstrapcdn.com/bootstrap/3.3.5/css/bootstrap.min.css">
  </head>

<body>
<div id="form">
  <div class="container">
    <div class="col-lg-6 col-lg-offset-3 col-md-6 col-md-offset-3 col-md-8 col-md-offset-
2">
      <div id="userform">
```


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```
<ul class="nav nav-tabs nav-justified" role="tablist">
  <li class="active"><a href="#signup" role="tab" data-toggle="tab">Sign up</a></li>
  <li><a href="#login" role="tab" data-toggle="tab">Log in</a></li>
</ul>
<div class="tab-content">
  <div class="tab-pane fade active in" id="signup">
    <h2 class="text-uppercase text-center"> Sign Up for Free</h2>
    <form id="signup">
      <div class="row">
        <div class="col-xs-12 col-sm-6">
          <div class="form-group">
            <label>First Name<span class="req">*</span> </label>
            <input type="text" class="form-control" id="first_name" required data-
validation-required-message="Please enter your name." autocomplete="off">
            <p class="help-block text-danger"></p>
          </div>
        </div>
        <div class="col-xs-12 col-sm-6">
          <div class="form-group">
            <label> Last Name<span class="req">*</span> </label>
            <input type="text" class="form-control" id="last_name" required data-
validation-required-message="Please enter your name." autocomplete="off">
            <p class="help-block text-danger"></p>
          </div>
        </div>
      </div>
      <div class="form-group">
        <label> Your Email<span class="req">*</span> </label>
        <input type="email" class="form-control" id="email" required data-validation-
required-message="Please enter your email address." autocomplete="off">
        <p class="help-block text-danger"></p>
      </div>
      <div class="form-group">
        <label> Your Phone<span class="req">*</span> </label>
        <input type="tel" class="form-control" id="phone" required data-validation-
required-message="Please enter your phone number." autocomplete="off">
        <p class="help-block text-danger"></p>
      </div>
    </form>
  </div>
</div>
```

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```
</div>
<div class="form-group">
  <label> Password<span class="req">*</span> </label>
    <input type="password" class="form-control" id="password" required data-
validation-required-message="Please enter your password" autocomplete="off">
    <p class="help-block text-danger"></p>
</div>
<div class="mrqn-30-top">
  <button type="submit" class="btn btn-larger btn-block"/>
  Sign up
</button>
</div>
</form>
</div>
<div class="tab-pane fade in" id="login">
  <h2 class="text-uppercase text-center"> Log in</h2>
  <form id="login">
    <div class="form-group">
      <label> Your Email<span class="req">*</span> </label>
        <input type="email" class="form-control" id="email" required data-validation-
required-message="Please enter your email address." autocomplete="off">
        <p class="help-block text-danger"></p>
    </div>
    <div class="form-group">
      <label> Password<span class="req">*</span> </label>
        <input type="password" class="form-control" id="password" required data-
validation-required-message="Please enter your password" autocomplete="off">
        <p class="help-block text-danger"></p>
    </div>
    <div class="mrqn-30-top">
      <button type="submit" class="btn btn-larger btn-block"/>
      Log in
    </button>
  </div>
</form>
</div>
</div>
```

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```
</div>
</div>
</div>
<!-- /.container -->
</div>
<script src="//code.jquery.com/jquery-1.11.3.min.js"></script>
<!-- Latest compiled and minified JavaScript -->
<script
src="https://maxcdn.bootstrapcdn.com/bootstrap/3.3.5/js/bootstrap.min.js"></script>
</body>
<!-- partial -->
<script src="./script.js"></script>

</body>
</html>
```

10.3 WEB

```
<html>

<style>
div.header{
  top: 0;
  position: fixed;
  padding-left: 400px;}
div.header1{
  top:20;
  position: fixed;
  padding-left: 490px;
}

*{
  margin:0;
  padding:0;
  border:0;
  outline:0;
  text-decoration:none;
  font-family:montserrat;
```

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```
}
```

```
body
```

```
{
```

```
background-image:url('{{url_for('static', background="oil-2_0-sixteen_nine.jpg')}}');
```

```
background-position: center top;
```

```
font-family:sans-serif;
```

```
background-size:cover;
```

```
margin-top:40px;
```

```
}
```

```
.main    input[type="text"],.main    input[type="text"],.main    input[type="text"],.main
```

```
input[type="text"],.main    input[type="text"],.main    input[type="text"],.main
```

```
input[type="text"]{
```

```
    border:0;
```

```
    background:none;
```

```
    display:block;
```

```
    margin:20px auto;
```

```
    text-align:center;
```

```
    border:2px solid black;
```

```
    padding:15px 3px;
```

```
    width:400px;
```

```
    outline:none;
```

```
    color:black;
```

```
    border-radius:0px;
```

```
    transition:0.25s;
```

```
    font-size:20;
```

```
}
```

```
.bor{
```

```
border:0;
```

```
    background:none;
```

```
    display:block;
```

```
    margin:20px auto;
```

```
    text-align:center;
```

```
    border:2px solid black;
```

```
    padding:10px 3px;
```

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```
width:500px;
outline:none;
color:black;
transition:0.25s;}
.main          input[type="text"]:focus,.main          input[type="text"]:focus,.main
input[type="text"]:focus,.main          input[type="text"]:focus,.main
input[type="text"]:focus,.main input[type="text"]:focus,.main input[type="text"]:focus{
width:280px;
color: black;
border-color:black;
}
.logbtn{
display:block;
width:35%;
height:50px;
border:none;
border-radius:24px;
background:linear-gradient(120deg,#3498db,#8e44ad,#3498db,#8e44ad);
background-size:200%;
color:black;
outline:none;
cursor:pointer;
transition:.5s;
font-size:25;
}
.logbtn:hover{
background-center;
}

input::placeholder{
color:purple;
font-family: verdana;
font-size: 15;
}
.bottom-text{
margin-top:60px;
text-align:center;
```

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font-size:13px;

}

</style>

<body>

<div class="navbar">

Contact |

Home

</div>

<center><h1 style="font-family:verdana;">CRUDE OIL PRICE
PREDICTOR</h1></center>

<form style="color: black;" class="main" action="/login" method="post">

<input type="text" name="year1" placeholder="Enter
previous 10th day price" style="color:black;"/>

<input type="text" name="year2" placeholder="Enter
previous 9th day price"/>

<input type="text" name="year3" placeholder="Enter
previous 8th day price"/>

<input type="text" name="year4" placeholder="Enter
previous 7th day price"/>

<input type="text" name="year5" placeholder="Enter
previous 6th day price"/>

<input type="text" name="year6" placeholder="Enter
previous 5th day price"/>

<input type="text" name="year7" placeholder="Enter
previous 4th day price"/>

<input type="text" name="year8" placeholder="Enter
previous 3th day price"/>

<input type="text" name="year9" placeholder="Enter
previous 2nd day price"/>

<input type="text" name="year10" placeholder="Enter

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```
previous 1st day price"/></font>
        <center><input type="submit" class="logbtn" value="Predict"></center>
        <div
            class="bor"><b><font
                color="white"
size=5>{{showcase}}</font></b></div>
        </form>
</div>
</body>

</html>
```

11 BUILD PYTHON CODE:

We have a built a flask file 'app.py' which is a web framework written in python for server-side scripting.

Let's see step by step procedure for building the backend application.

The app starts running when the "__name__" constructor is called in main.

render_template is used to return HTML files.

"GET" method is used to take input from the user.

"POST" method is used to display the output to the user.

Importing Libraries

Routing to the html Page

For predicting next day's crude oil prices we consider n_steps=10.

the input for prediction, index starting from the date 10 days before the first date in the test dataset.

Then, reshape the inputs to have only 1 column and predict using model_predict predefined function.

```
import numpy as np # used for numerical analysis
from flask import Flask, render_template, request, url_for, redirect # Flask is
a application used to run/serve our application
# request is used to access the file which is uploaded by the user in our
application
# render_template is used for rendering the html pages
from tensorflow.keras.models import load_model # we are loading our model from
keras
```

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```
app = Flask(__name__) # our flask app
model = load_model('crude_oil_price_prediction.h5') # loading the model in the flask app

@app.route('/', methods=['GET', 'POST'])
def home():
    error = None
    if request.method == 'POST':
        if request.form['username'] != "PNT2022TMID43400" or request.form['password'] != "7155":
            error = 'Invalid Credentials. Please try again.'
        else:
            return redirect(url_for('mains'))
    return render_template('login.html', error=error)

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

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

@app.route('/about')
def home1():
    return render_template("index.html") # rendering html template

@app.route('/predict')
def home2():
    return render_template("web.html") # rendering html template

@app.route('/contact')
def contact():
    return render_template("contact.html")

@app.route('/login', methods=['POST']) # route for our prediction
def login():
    a = request.form['year1']
    b = request.form['year2']
```


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```
c = request.form['year3']
d = request.form['year4']
e = request.form['year5']
f = request.form['year6']
g = request.form['year7']
h = request.form['year8']
i = request.form['year9']
j = request.form['year10'] # requesting the file
x_input = [[float(a), float(b), float(c), float(d), float(e), float(f), float(g),
float(h), float(i), float(j)]]
print(x_input)
lst_output = model.predict(x_input)
lst_output = np.round(lst_output[0][0], 2)
return render_template("web.html", showcase='The Predicted crude oil price
is : Rs. '+str(lst_output))

if __name__ == '__main__':
    app.run(debug=False)
```

12 RUN THE APP IN LOCAL BROWSER:

Open anaconda prompt from the start menu

Navigate to the folder where your python script is.

Now type “python app.py” command

Navigate to the localhost where you can view your webpage

12.1 SHOWCASTING PREDICTION ON UI:

In home page we get the summary of the project.

For the time series approach the user has to give the past 10 days price of crude oil to predict the future crude oil prediction.

If we give input of past 10 days and click the predict for the next day.

The output will be displayed to the user interface.

13 TRAIN THE MODEL ON IBM

Build a Machine Learning Model and deploy it on the IBM Cloud.

13.1 REGISTER FOR IBM CLOUD:

create an IBM cloud account to login and also for train the model.

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13.2 TRAIN THE ML MODEL ON IBM:

Train the machine learning model on the IBM watson.

13.3 INTEGRATE FLASK WITH SCORING END POINT:

Integration the scoring endpoint of the flask.

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14 RESULT

A Machine Learning model using LSTM was developed and a user interface was created.