LABORATORY REPORT

Application Development Lab (CS33002)

B.Tech Program in ECSc

Submitted By

Name: Shreyaa Venkateswaran

Roll No: 2230120



Kalinga Institute of Industrial Technology (Deemed to be University) Bhubaneswar, India

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Experiment Number	3
Experiment Title Regression Analysis for Stock Prediction	
Date of Experiment	21-01-2025
Date of Submission	27-01-2025

1. Objective:

To perform stock price prediction using Linear Regression and LSTM models.

2. Procedure:

- 1. Collect historical stock price data.
- 2. Preprocess the data for analysis (missing data, scaling, splitting into train/test).
- 3. Implement Linear Regression to predict future stock prices.
- 4. Design and train an LSTM model for time-series prediction.
- 5. Compare the accuracy of both models.
- 6. Create a Flask backend for model predictions.
- 7. Build a frontend to visualize predictions using charts and graphs.

3. Code:

Model Training Code:

import numpy as np

import pandas as pd

import matplotlib.pyplot as plt #to plot graphs

import seaborn as sns #to plot graphs

from sklearn.linear_model import LinearRegression #for linear regression model

```
sns.set() #setting seaborn as default
import math
import warnings
warnings.filterwarnings('ignore')
from google.colab import drive
drive.mount('/content/drive')
data=pd.read csv("/content/drive/MyDrive/Stock/dataset/INDIAVI
X.csv") #reads the input data
data.head() #displays the first five rows
data.info()
data.describe(include ='all')
data.isnull().sum()
data.head()
sns.pairplot(data)
plt.show()
x=data[['Date','Open','High','Low','Close','Previous','Change','%Cha
nge']].values
y=data[['Close']].values
from sklearn.model selection import train test split
#split to train and test data
                                                 train test split(x,y,
x train,
           x test,
                      y train,
                                  y test
test size=0.2,random state=0)
data['Date'] = pd.to datetime(data['Date'])
# Extract year, month, day, day of the week from the Date
data['Year'] = data['Date'].dt.year
data['Month'] = data['Date'].dt.month
```

```
data['Day'] = data['Date'].dt.day
data['Day of week'] = data['Date'].dt.dayofweek # Monday=0,
Sunday=6
X = data[['Open', 'High', 'Low', 'Previous', 'Change', '%Change',
'Year', 'Month', 'Day', 'Day of week']]
X train,
           X test,
                   y train, y test
                                       = train test split(X,
test size=0.2, random state=42)
print(f"Training data size: {len(X train)}")
print(f"Test data size: {len(X test)}")
#using linear regression
lm=LinearRegression()
lm.fit(X train,y train)
lm.coef
#values from 0 to 1
#0 model explain None of the variability
#1 model explain Entire of the variability
lm.score(X train,y train)
from sklearn.impute import SimpleImputer
# Create an imputer instance
imputer = SimpleImputer(strategy='mean') # Use 'median' or other
strategies if more appropriate
# Fit and transform the training and test data
X train = imputer.fit transform(X train)
X \text{ test} = \text{imputer.transform}(X \text{ test})
#predict the output(predictions) using the test data
predictions = lm.predict(X test)
```

```
from sklearn.metrics import r2_score
r2_score(y_test, predictions)
#load actual and predecited values side by side
dframe=pd.DataFrame({'actual':y test.flatten(),'Predicted':predictio
ns.flatten()})
#flatten toget single axis of data (1 dimension only)
dframe.head(15)
graph = dframe.head(10)
graph.plot(kind='bar')
plt.title('Actual vs Predicted')
plt.ylabel('Closing price')
#using scatter plot compare the actual and predicted data
fig = plt.figure()
plt.scatter(y test,predictions)
plt.title('Actual versus Prediction')
plt.xlabel('Actual', fontsize=20)
plt.ylabel('Predicted', fontsize=20)
import matplotlib.pyplot as plt
# Calculate residuals
residuals = y_test - predictions
# Create the residual plot
plt.figure(figsize=(8, 6))
plt.scatter(predictions, residuals, alpha=0.7)
plt.axhline(0, color='red', linestyle='--', linewidth=1) # Horizontal
line at residual = 0
plt.title("Residual Plot", fontsize=16)
```

```
plt.xlabel("Predicted Values", fontsize=14)
plt.ylabel("Residuals (Actual - Predicted)", fontsize=14)
plt.grid(alpha=0.3)
plt.show()
import math
from sklearn import metrics
#metrics to find accuracy of continous variables
print('Mean
                                 Abs
                                                           value:'
,metrics.mean absolute error(y test,predictions))
print('Mean
                                                          squared
value:',metrics.mean squared error(y test,predictions))
print('root
                      mean
                                         squared
                                                             error
value:',math.sqrt(metrics.mean squared error(y test,predictions)))
import os
folder path = '/content/drive/My Drive/your folder name'
os.makedirs(folder path, exist ok=True) # This creates the folder
if it doesn't exist
# Save the model
model path=os.path.join(folder path,'linear regression model.pkl)
dump(lm, model path)
print(f"Model saved to {model path}")
index.html:
<!DOCTYPE html>
<html lang="en">
```

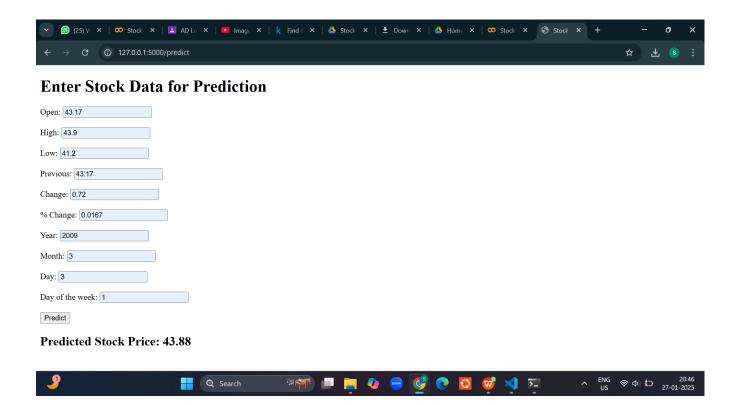
```
<head>
<meta charset="UTF-8">
           name="viewport"
                                 content="width=device-width,
<meta
initial-scale=1.0">
<title>Stock Price Prediction</title>
</head>
<body>
<h1>Enter Stock Data for Prediction</h1>
<form action="/predict" method="post">
<label for="Open">Open:</label>
<input type="text" name="Open" required><br><br>
<label for="High">High:</label>
<input type="text" name="High" required><br><br>
<label for="Low">Low:</label>
<input type="text" name="Low" required><br><br>
<label for="Previous">Previous:</label>
<input type="text" name="Previous" required><br><br>
<label for="Change">Change:</label>
<input type="text" name="Change" required><br><br>
<label for="%Change">% Change:</label>
<input type="text" name="%Change" required><br><br>
<label for="Year">Year:</label>
<input type="text" name="Year" required><br><br>
<label for="Month">Month:</label>
<input type="text" name="Month" required><br><br>
```

```
<label for="Day">Day:</label>
<input type="text" name="Day" required><br><br>
<label for="Day of week">Day of the week:</label>
<input type="text" name="Day_of_week" required><br><br>
<button type="submit">Predict</button>
</form>
<h2>{{ prediction text }}</h2>
</body>
</html>
app.py (Flask code):
from flask import Flask, render template, request
from joblib import load
import numpy as np
# Initialize Flask app
app = Flask( name )
# Load the trained model
model path = 'model/linear regression model.pkl' # Path to model
in your folder
model = load(model path)
@app.route('/')
def home():
return render template('index.html')
@app.route('/predict', methods=['POST'])
def predict():
```

```
try:
# Get the form data (input values from the user)
features = [
float(request.form['Open']),
float(request.form['High']),
float(request.form['Low']),
float(request.form['Previous']),
float(request.form['Change']),
float(request.form['%Change']),
int(request.form['Year']),
int(request.form['Month']),
int(request.form['Day']),
int(request.form['Day of week'])
1
# Convert features to a 2D array (model expects 2D array)
features = np.array(features).reshape(1, -1)
# Predict the stock price using the trained model
prediction = model.predict(features)
# Ensure the prediction is a scalar value
prediction value = prediction.item() # This converts the numpy
array to a scalar
# Return the predicted value to the user
         render template('index.html',
                                        prediction text='Predicted
Stock Price: {:.2f}'.format(prediction value))
except Exception as e:
```

```
return render_template('index.html', prediction_text="Error: " +
str(e))
if __name__ == '__main__':
app.run(debug=True)
```

4. Results/Output:



5. Remarks:

Built a linear regression model for stock data spanning over a few years. The frontend is done using HTML while the backend is done using flask. The model shows pretty accurate results.

Initially there was a problem in integrating the flask interface along with the model however it was later rectified.

Website link: Stock Price Prediction

GitHub link: https://github.com/ShreyaaVenkateswaran

Shreyaa Venkateswaran	Signature of the Lab Coordinator