# Stock Market Forecasting Using LSTM

# **Project Overview**

This project presents a **Streamlit web application** that utilizes **Long Short-Term Memory (LSTM)** neural networks to predict stock market prices based on historical data. The application allows users to:

- Upload stock data
- Visualize price trends
- Train an LSTM model
- Evaluate performance
- Predict future stock prices based on user inputs

#### **Features**

- Data Upload: Users can upload a CSV file containing stock market data.
- Data Visualization: The application plots the normalized closing prices over time.
- LSTM Model Training: Enables training on uploaded data to learn patterns.
- Performance Metrics: Shows MAE and RMSE for model evaluation.

• Future Price Prediction: Accepts statistics like previous close, open, high, low, and volume for prediction.

### Installation

1. Clone the Repository:

```
git clone https://github.com/Tesa2035/Stock-Market-Forecasting3.git
```

2. Navigate to the Project Directory:

```
cd Stock-Market-Forecasting3
```

3. Create Virtual Environment:

```
python3 -m venv venv
source venv/bin/activate  # Windows: venv\Scripts\activate
```

4. Install Requirements:

```
pip install -r requirements.txt
```

### Usage

Run the Application

streamlit run app.py

#### Interact with the App

- Upload Data: Ensure the file contains columns: date, open, high, low, close, adjclose, volume.
- Visualize Data: Normalized closing prices are plotted.
- Train Model: Automatically preprocesses and trains the LSTM model.
- Predict Prices: Enter relevant stock metrics to forecast next price.

## **Data Preparation**

- Format: CSV with columns date, open, high, low, close, adjclose, volume.
- Date Format: Ensure date is in YYYY-MM-DD.
- Missing Values: Clean the data before upload.
- Frequency: Use daily time series data.

#### Model Details

- Architecture:
  - 2 LSTM layers with 50 units
  - Dense layer with 25 units
  - Dense output layer with 1 unit
- Sequence Length: Uses past 10 days to predict the next closing price.
- Scaling: All features are Min-Max scaled to range [0, 1].

# **Screenshots**

- Figure 1: Normalized Closing Price Over Time
- Figure 2: LSTM Model Training Progress
- Figure 3: Future Price Prediction Interface

# Repository Link

https://github.com/Tesa2035/Stock-Market-Forecasting3