

Stock Market Price Prediction Project

Using LSTM Machine Learning Model to Navigate the Complexities of Stock Predictions for Investors and Analysts.



Hacker X Hacker

AIML - 1



TEAM OVERVIEW

Stock Market Price Prediction Project

Using LSTM Machine Learning Model



Abhiyanshu Anand 2200911530008

Project Lead

Experienced in machine learning and data analysis, overseeing the project lifecycle and ensuring alignment with business objectives.

01



Abhishek Rajput 2200911530005

Data Scientist

Specializes in LSTM models and time series forecasting, responsible for developing predictive algorithms and analyzing stock market trends.

02



Abhishek Singh 2200911530006

Software Engineer

Focuses on building the software infrastructure for data processing and model deployment, ensuring scalability and performance.

03



Abhishek Upadhayay 2200911530007

Business Analyst

Analyzes market data and client requirements, bridging the gap between technical and business teams to drive project success.

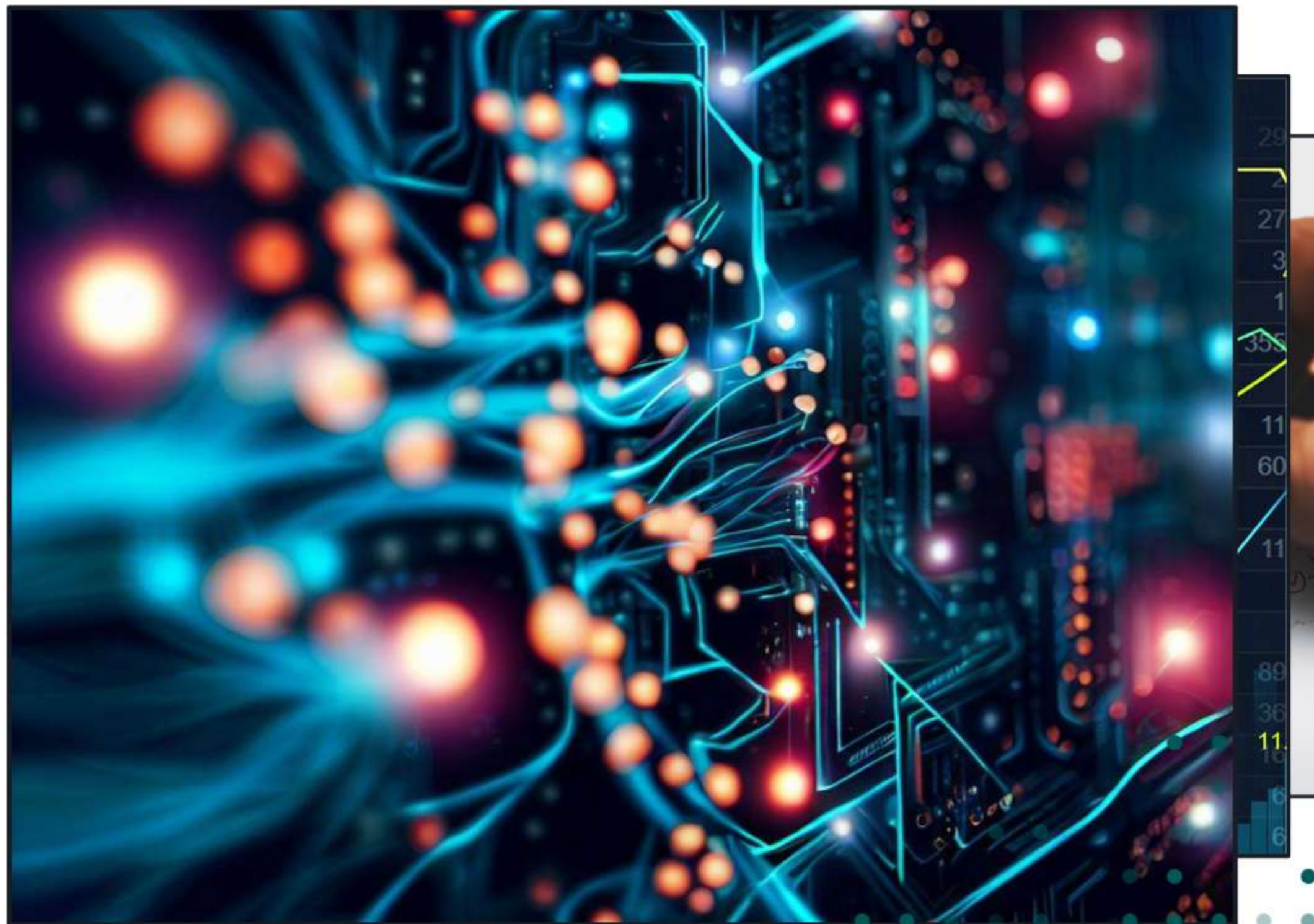
04



01

Goal of Project

To predict the next day's stock price using an LSTM model, leveraging real-time data.



02

Real-Time Data Fetching

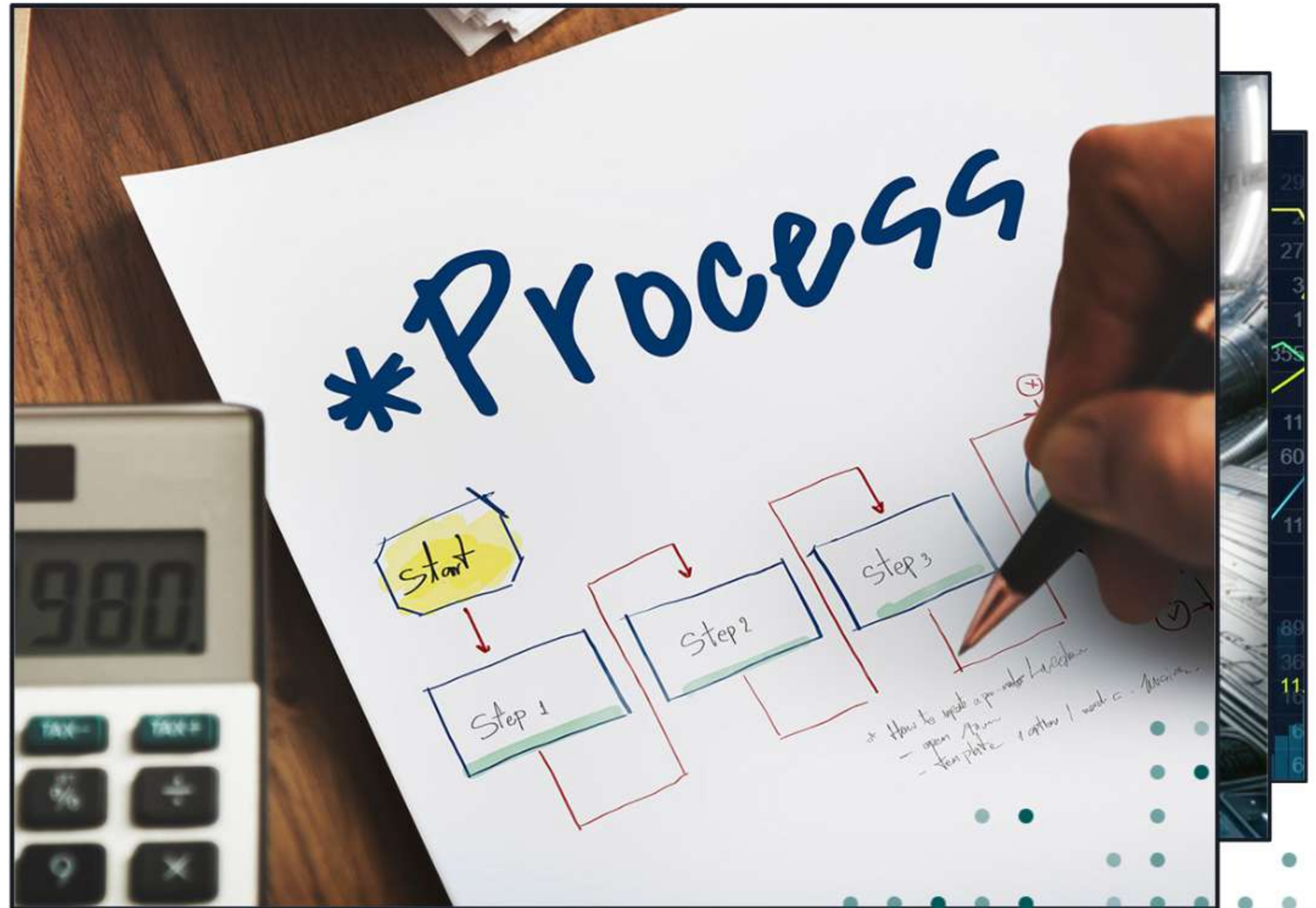
- The project involves fetching real-time stock data, ensuring accuracy in predictions.



03

Data Preprocessing

Preprocessing techniques will be applied to prepare the data for effective model training.



04

Dynamic Predictions

- Predictions are made dynamically through a Flask web application, enhancing user interaction.



05

Importance of Stock Market Prediction

Predicting stock prices is crucial for investors and businesses seeking strategic advantages.



06

Real-World Applications

- This project addresses real-world financial decision-making processes, providing valuable insights.



07

Combining Skills

The project integrates machine learning and data science, showcasing practical applications of these fields.



TEAM CONTRIBUTIONS

Project Team Overview

Meet the talented individuals driving our innovative project



01 **Abhiyanshu Anand** | LSTM Model Developer

Focused on the training and fine-tuning of the LSTM model, I built and integrated the prediction model that forms the backbone of our application.



02 **Abhishek Singh** | Web Application Developer

Contributed to the development of the web application using Flask. Integrated the machine learning model into the app and crafted an intuitive frontend for user engagement.



03 **Abhishek Rajput** | Data Engineer

Oversaw data preprocessing and feature engineering. Utilized the yfinance library to fetch real-time stock data, ensuring it was cleaned and scaled appropriately for analysis.



04 **Abhishek Upadhayay** | Documentation Specialist

Responsible for comprehensive project documentation and presentation preparation, ensuring that all technical details and results were effectively communicated to stakeholders.



2015-01-01 to
Present

Data Preprocessing

The Close prices are normalized using MinMaxScaler, scaling the data to a range between 0 and 1 to enhance model performance.

Training Period

Prediction

After training, the model can predict the next day's stock price for any stock ticker input by the user, providing actionable insights.

Data Collection

The application fetches stock data using the yfinance library, retrieving Open, High, Low, and Close prices from Yahoo Finance since January 1, 2015.

Preprocessing Phase

LSTM Model Training

An LSTM neural network is trained on the previous 60 days of stock prices, enabling it to learn patterns for predicting the next day's stock price.

Post-Training

STOCK PREDICTION STEPS

Stock Price Prediction Process

A Step-by-Step Overview

Data Preprocessing

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Preprocessing Phase

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Deployment Phase

Flask Web Application

A Flask web application is implemented, allowing users to input a stock ticker (e.g., AAPL) and receive the predicted stock price for the next day.

STOCK PREDICTION STEPS

Stock Price Prediction Process

A Step-by-Step Overview

FLASK APP DEMO

Demonstrating the Flask App

Steps and Tips for Live Demo

01



Open Flask App

Start the Flask application by executing the command **python app.py** in your terminal to initiate the server.

02



Access Web Interface

Navigate to **http://127.0.0.1:5000/** in your web browser to access the application's user interface.

03



Input Stock Ticker

In the input field, enter a stock ticker symbol, such as AAPL, and click the 'Predict' button to generate a prediction.

04



View Prediction

Observe the predicted stock price displayed dynamically on the webpage after you click 'Predict'.

TECH STACK OVERVIEW

Technologies Used

An Overview of Programming Languages, Libraries, and Machine Learning Concepts



Programming Languages

Python serves as the primary language for data processing, model training, and building the Flask web application, providing a robust and versatile foundation for development.



Libraries and Frameworks

Utilizing key libraries enhances functionality: yfinance for real-time stock data, scikit-learn for data preprocessing, TensorFlow/Keras for LSTM model development, and Flask for web deployment.



Machine Learning Concepts

Understanding LSTM (Long Short-Term Memory) networks is crucial for handling sequential data like stock prices, while normalization is essential for improving model performance.

Challenges and Solutions in Stock Data Management

Addressing Key Issues in Real-Time Data Processing



01



Handling Dynamic Data

- Integrated yfinance to dynamically fetch the latest stock data, eliminating manual updates. The app retrieves real-time data every time a prediction is made, ensuring users have the most current information.

02



Ensuring Accurate Predictions

Utilized an LSTM model tailored for sequential data, trained on the past 60 days of stock information. This approach captures essential trends and patterns, enhancing the reliability of stock market predictions.

03



Handling JSON Serialization Errors

Implemented a solution for converting NumPy float32 values to Python-native floats, ensuring compatibility with JSON serialization. This step resolves potential data transfer issues, facilitating smooth integration.

App Results and Future Scope

Key Findings and Enhancements



Accurate Predictions

The app effectively predicts the next day's stock price for any valid stock ticker by utilizing real-time data that is fetched dynamically, ensuring users receive timely and relevant information.



User-Friendly Interface

The Flask web app is designed to facilitate easy user interaction, allowing individuals to input a stock ticker and receive immediate predictions, making it accessible even for those with minimal technical knowledge.



Enhancing Predictive Accuracy

Future updates may include additional features like trading volume and sentiment analysis derived from news articles, which could significantly improve the model's predictive accuracy and provide more nuanced insights.



Visualizing Data

The application can be further enhanced by integrating improved visualization tools, such as displaying stock price graphs and historical trends, enabling users to analyze stock performance over time.



Model Reliability

By implementing cross-validation techniques, we can enhance the reliability of the predictive model, ensuring that the forecasts provided by the app are more robust and trustworthy.

Conclusion and Questions

Key Insights and Interactive Discussion



01 Project Integration

This project effectively merges machine learning techniques with real-time data processing capabilities, demonstrating a modern approach to stock price analysis.

02 Prediction Tool

We developed a practical stock price prediction tool that leverages complex algorithms to anticipate market movements based on historical data.

03 LSTM Model Utilization

The model employs Long Short-Term Memory (LSTM) architecture, which is particularly well-suited for time series forecasting, enhancing prediction accuracy.

04 Engagement Invitation

We encourage you to pose any questions related to the project, the underlying model, or the web application interface, fostering an interactive discussion.



Transform Your Investment Strategy

Join us in exploring how predictive analytics and machine learning can revolutionize your stock trading and

- investment strategies today. Don't miss out on the future
- of trading

