Comprehensive EDA Report - Task 4

1. Data Overview

The dataset contains historical stock price data with the following key columns:

• Date: The trading date

• Open: Opening stock price

• **High**: Highest stock price during the trading session

• Low: Lowest stock price during the trading session

• Close: Closing stock price (target variable for prediction)

• Volume: Number of shares traded

Data Sample

A snapshot of the dataset:

	Α	В	C	D	Е	F	G	Н
1		Date	Adj Close	Close	High	Low	Open	Volume
2	0	3/17/1980	2.296798	3.291227	3.344743	3.291227	0	41109
3	1	3/18/1980	2.306134	3.304606	3.358122	3.304606	0	9343
4	2	3/19/1980	2.306134	3.304606	3.304606	3.304606	3.304606	0
5	3	3/20/1980	2.306134	3.304606	3.358122	3.304606	0	10277
6	4	3/21/1980	2.362154	3.38488	3.438396	3.38488	0	8409
7	5	3/24/1980	2.37149	3.398259	3.451775	3.398259	0	2803
8	6	3/25/1980	2.352816	3.371501	3.425017	3.371501	0	27095
9	7	3/26/1980	2.352816	3.371501	3.425017	3.371501	0	1869
10	8	3/27/1980	2.324807	3.331364	3.38488	3.331364	0	14015
11	Q	3/28/1980	2 31547	2 217925	3 371501	2 217925	n	36438

2. Data Cleaning and Preprocessing

- Missing Values: Checked for missing values and filled them using forward fill.
- Date Conversion: Converted 'Date' column to a datetime format and set it as the index.
- Feature Scaling: Applied MinMaxScaler (0,1) for normalizing data.

3. Data Visualization

3.1 Stock Price Trend Over Time

A line plot of the closing prices shows trends and seasonality.

3.2 Moving Averages

- **50-day Moving Average**: Captures mid-term trends.
- 200-day Moving Average: Identifies long-term trends.

3.3 Volume Analysis

 Volume spikes correspond to sharp price movements, often signaling market activity.

4. Trend, Seasonality, and Anomalies

4.1 Trend Analysis

Using rolling averages, we identify an overall upward/downward trend in stock prices.

4.2 Seasonality Detection

Using decomposition techniques, we detect cyclic patterns in stock price behavior.

4.3 Anomaly Detection

By using boxplots and Z-score analysis, we identify outliers in price fluctuations.

5. Feature Selection Justification

Selected Features:

- 1. Close Price (Target Variable)
- 2. Lagged Features: Previous closing prices are used to model temporal dependencies.
- 3. Rolling Statistics: Moving averages (e.g., 10-day, 50-day) improve trend analysis.
- 4. Trading Volume: High volume often precedes significant price changes.

6. Data Preprocessing Decisions

- **Normalization**: Applied MinMax scaling to standardize input values.
- Sequence Creation: Transformed data into sequences of past 'n' days for LSTM input.
- Train-Test Split: 80% training, 20% testing.

Conclusion

The dataset exhibits strong trends, periodic seasonality, and volume-driven price changes. The selected features balance trend-following characteristics with anomaly detection.