Exploratory Data Analysis Report

1. Data Overview

• The dataset contains historical stock price data from 1980 onward.

Key attributes:

- o Date: The trading day.
- o Open: Opening price of the stock.
- High: Highest price reached during the trading day.
- Low: Lowest price reached during the trading day.
- o Close: Closing price of the stock.
- o Adj Close: Adjusted closing price (accounts for stock splits & dividends).
- o Volume: Number of shares traded.

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Initial Observations:

- The dataset has missing values in key columns, which need to be addressed.
- There are periods of high and low volatility, indicating fluctuating market conditions.

2. Handling Missing Data

Identified Issues:

- Missing values in Close, Open, High, Low, and Volume.
- Some Open values are recorded as θ, which might indicate missing data.

• Solution Applied:

- Forward-fill method was used to propagate the last valid price value to maintain continuity.
- Dropped rows where Date was missing to preserve time consistency.
- Ensured all processed data is consistent for feature engineering.

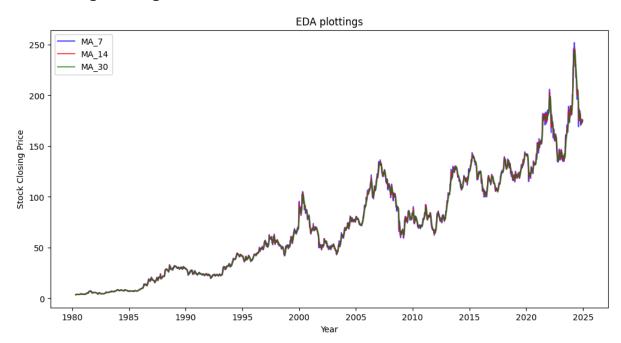
3. Key Visualizations & Relationships

3.1 Stock Price Trends Over Time



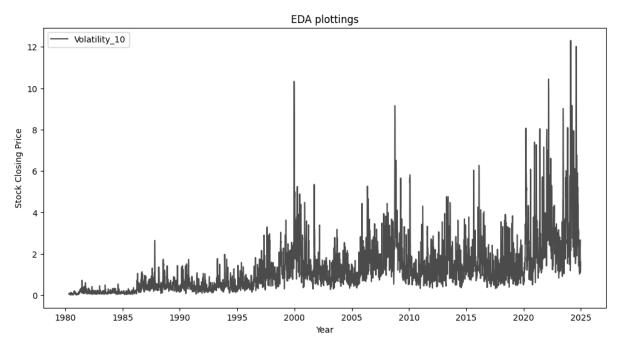
- **Time series plot** of closing prices shows a long-term **upward trend** with periodic fluctuations.
- Certain periods exhibit sharp increases/decreases, possibly linked to market events.

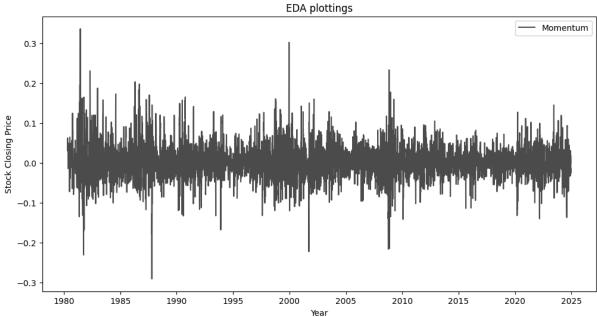
3.2 Moving Averages for Trend Identification



- Applied 7-day, 14-day, and 30-day Moving Averages to smooth price fluctuations.
- Moving averages help in identifying short-term and long-term trends.

3.3 Seasonality & Volatility Analysis





- Stock prices are cyclical, with periods of high volatility followed by stabilization.
- Rolling 10-day standard deviation used to track volatility over time.

4. Trends, Seasonality, and Anomalies

4.1 Long-Term Trend

 Overall, the stock price has shown a strong upward trend, driven by growth in market capitalization and investor interest.

4.2 Seasonality

- The stock market follows cycles, where prices tend to be higher during certain periods (e.g., Q4 earnings seasons, holiday spikes).
- Yearly seasonality patterns observed: Stronger gains at the start of the year, potential dips mid-year.

4.3 Volatility & Anomalies

- Certain **sharp price drops/spikes** are **not explained by normal trends** (potential market crashes, economic shifts).
- Periods of **high volatility** coincide with significant market events (e.g., financial crises, policy changes).

5. Feature Selection: Justification & Choices

The following features were selected based on **historical stock market trends and financial indicators**:

5.1 Moving Averages (MA_7, MA_14, MA_30)

- Purpose: Smoothens price fluctuations and identifies trends.
- **Justification**: Short-term traders use 7-day MA, while long-term investors track 30-day MA.

5.2 Volatility Indicator (Rolling 10-day Standard Deviation)

- **Purpose**: Measures price fluctuation intensity.
- **Justification**: High volatility often signals market uncertainty or upcoming trend reversals.

5.3 Momentum Indicator (5-day Rate of Change)

- Purpose: Measures price movement speed.
- **Justification**: If momentum is high, the stock is experiencing strong buying/selling pressure.

5.4 Relative Strength Index (RSI_14)

- Purpose: Determines if a stock is overbought (>70) or oversold (<30).
- Justification: Helps detect trend reversals by identifying extreme conditions.

6. Data Preprocessing & Normalization

6.1 Normalization Approach

- Used StandardScaler to normalize numerical features (Close, MA_7, MA_14, Momentum, etc.).
- Why?
 - o Ensures all features are on a **similar scale** to prevent model bias.
 - Helps in better model convergence for regression-based approaches.

6.2 Final Processed Dataset

- No missing values after preprocessing.
- All numerical features standardized.
- The dataset is now **clean and ready** for model training.

Final Takeaways

- Stock price follows trends with clear seasonal cycles.
- Volatility & anomalies indicate significant market events.
- Feature selection is based on financial indicators commonly used by traders.
- Processed dataset is well-structured for predictive modeling.