# STOCK MARKET ANALYSIS

## 1. Objectives

- **Goal**: Predict the closing price of a company's stock for a specific future date.
- **Scope**: Determine the range of dates and the accuracy requirements for the forecast.

#### 2. Data Collection

- **Historical Data**: Gather historical stock price data, including closing prices, for the company in question. This data might also include open, high, low prices, volume, and other relevant metrics.
- **Sources**: Utilize financial databases and APIs such as Yahoo Finance, Here we selected NVDA company stock price data from Yahoo Finance

## 3. Data Preparation

- Cleaning: Handle missing values, remove duplicates, and filter out irrelevant data.
- **Feature Engineering**: Create additional features like moving averages, trading volume trends, and technical indicators.

## 4. Exploratory Data Analysis (EDA)

### Data cleaning-

- o **Checking Invalid Records:** Identify and rectify any invalid entries in the dataset.
- o **Missing Value Detection and Imputation:** Detect missing values and apply appropriate imputation methods to fill gaps.
- o **Duplicated Records:** Find and remove duplicate records to ensure data integrity.
- o **Handling Outliers:** Detect and handle outliers that may skew the analysis and model performance.

#### Numerical Data visualizations-

- o **Pie Chart:** Visualize the distribution of categorical features such as book genres.
- o **Bar Graph:** Display the frequency of top authors or the number of books published each year.
- o **Histogram:** Show the distribution of numerical features such as book ratings.

- o **Scatterplot:** Explore relationships between two numerical features, such as ratings and number of pages.
- o **Heatmap:** Display correlations between different features in the dataset.
- o **Boxplot:** Outlier detection using boxplot

### 5. Model Selection

Choose a forecasting model based on the complexity and requirements of your project:

### • Time Series Models:

- ARIMA (AutoRegressive Integrated Moving Average): Useful for univariate time series data.
- SARIMA (Seasonal ARIMA): Extends ARIMA to account for seasonality.
- Exponential Smoothing: Includes models like Holt-Winters for capturing trend and seasonality.

## Machine Learning Models:

- Linear Regression: Simple model for predicting prices based on features.
- Decision Trees/Random Forests: For handling non-linear relationships.
- Support Vector Machines (SVM): Useful for regression tasks with complex relationships.

### Deep Learning Models:

- Recurrent Neural Networks (RNNs): Including Long Short-Term Memory (LSTM) networks, which are effective for sequential data.
- Gated Recurrent Units (GRUs): Similar to LSTMs but with fewer parameters.

## 6. Model Training

- **Split Data**: Divide data into training and testing sets. Consider using a rolling window approach for time series data.
- Hyperparameter Tuning: Optimize model parameters to improve performance.
- Cross-Validation: Use techniques like k-fold cross-validation to assess model performance.

### 7. Model Evaluation

Metrics: Evaluate models using metrics like Mean Absolute Error (MAE),
Root Mean Squared Error (RMSE), and R-squared.

# 8. Deployment

- **Streamlit App:** Develop a Streamlit application to deploy the Stock Market Analysis System.
- **User Interface**: Develop an interface (web or desktop) where users can input a date and receive a forecasted stock price.