## **Stock Price Prediction using LSTM Neural Networks**

#### **Project Overview**

This project fetches, processes, visualizes, and builds a machine learning model for predicting Google's (or any other) stock prices using historical data from Yahoo Finance.

## **Step-by-Step Execution**

#### 1. Data Collection

- Import required libraries (pandas, numpy, yfinance, etc.).
- Define a date range (last 20 years).
- Download Google's stock data using yf.download().

## 2. Exploratory Data Analysis (EDA)

- Display dataset structure using .head(), .shape(), .describe(), and .info().
- Visualize the closing stock prices using matplotlib.
- Use a custom function plt\_graph() to plot different stock attributes dynamically.

## **Machine Learning Model (LSTM)**

# 3. Model Preparation

- Import TensorFlow and Keras to build an LSTM-based neural network.
- Define a Sequential model containing LSTM layers to capture time dependencies.
- Save the trained model as "stock\_price\_model.keras".

## 4. Working Principle

- The model learns historical stock price patterns using LSTMs, which are suited for sequential data.
- It processes past stock values, identifies trends, and predicts future prices.
- The final trained model is saved for future use.

#### **Model Breakdown & Working**

## 5. Model Architecture (Sequential)

- Sequential(): Initializes a sequential model.
- LSTM(128, return\_sequences=True): First LSTM layer with 128 units; returns sequences for further learning.

- LSTM(64, return\_sequences=False): Second LSTM layer with 64 units, outputting a final sequence.
- Dense(25): Fully connected layer with 25 neurons to process LSTM outputs.
- Dense(1): Final Dense layer with a single neuron for predicting the next stock price.

# 6. Compilation & Training

- model.compile(optimizer='adam', loss='mean\_squared\_error'):
  - o Uses the Adam optimizer for efficient learning.
  - o Applies Mean Squared Error (MSE) loss function, best suited for regression tasks.
- model.fit(x\_train, y\_train, batch\_size=16, epochs=10):
  - Trains the model on past stock prices (x\_train, y\_train).
  - Uses a batch size of 16 for 10 training epochs.

## 7. Prediction & Application

- predictions = model.predict(x\_test):
  - The trained model forecasts stock prices based on test data (x\_test).
  - Helps analyze past trends to predict future stock prices.

## Conclusion

This project effectively combines data analysis and deep learning to predict stock prices using LSTM networks. It leverages historical stock data, visualizes key trends, and builds a robust prediction model using TensorFlow/Keras. The trained model can be used for future predictions, making it valuable for stock market analysis.