

# Stock Price Movement Analysis Using Python

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## 1. Introduction

Stock price analysis is crucial for traders and investors to make informed decisions. This report explores short-term stock price movements using Python and various libraries such as pandas, numpy, matplotlib, seaborn, and mplfinance. The focus is on visualizing trends, calculating indicators, and analyzing volatility.

## 2. Methodology

### 2.1 Data Collection

Stock data is loaded from a CSV file containing historical prices. Key columns include Close Price, Moving Averages, Daily Returns, Log Returns, and Volatility.

### 2.2 Data Processing & Analysis

Moving Averages (SMA\_10 & SMA\_20): Used to smooth out price fluctuations.

Daily & Log Returns: Measures stock price changes over time.

Volatility: Captures market fluctuations.

Candlestick Charts: Visual representation of stock price movements.

### 2.3 Libraries Used

pandas: Data handling & processing.

numpy: Numerical calculations.

matplotlib & seaborn: Data visualization.

mplfinance: Candlestick charting.

## 3. Code Implementation

```
# Import necessary libraries
```

```
import pandas as pd # Data handling and processing
```

```
import numpy as np # Numerical operations
```

```
import matplotlib.pyplot as plt # Plotting graphs
```

```
import seaborn as sns # Statistical data visualization
```

```
import mplfinance as mpf # Candlestick charting
```

```
# Load stock data from CSV file
```

```
stock_data = pd.read_csv("stock_data.csv", parse_dates=["Date"], index_col="Date") # Load
```

and parse dates

```
# Filter for a specific stock symbol (e.g., "UTIBANK")
```

```
symbol = "UTIBANK"
```

```
stock_data = stock_data[stock_data["Symbol"] == symbol] # Filter the dataset for the  
chosen symbol
```

```
# --- STOCK PRICE MOVEMENT ---
```

```
plt.figure(figsize=(12, 6)) # Set figure size for better readability
```

```
plt.plot(stock_data["Close"], label="Closing Price", color="blue") # Plot closing price in blue
```

```
plt.plot(stock_data["SMA_10"], label="10-Day SMA", color="red", linestyle="dashed") # 10-  
day SMA in red dashed
```

```
plt.plot(stock_data["SMA_20"], label="20-Day SMA", color="green", linestyle="dashed") #  
20-day SMA in green dashed
```

```
plt.title(f"{symbol} Stock Price Movement") # Title of the plot
```

```
plt.xlabel("Date") # Label for x-axis
```

```
plt.ylabel("Price (INR)") # Label for y-axis
```

```
plt.legend() # Display legend
```

```
plt.grid() # Show grid for better visualization
```

```
plt.show() # Display the plot
```

```
# --- DAILY RETURNS ---
```

```
plt.figure(figsize=(12, 6)) # Set figure size
```

```
plt.plot(stock_data["Daily_Return"], label="Daily Return", color="purple") # Plot daily  
returns in purple
```

```
plt.axhline(0, linestyle="--", color="black", linewidth=0.8) # Zero-line reference for better  
comparison
```

```
plt.title(f"{symbol} Daily Returns") # Title of the plot
```

```
plt.xlabel("Date") # Label for x-axis
```

```
plt.ylabel("Return") # Label for y-axis
```

```
plt.legend() # Display legend
```

```
plt.grid() # Show grid for better visualization
```

```
plt.show() # Display the plot
```

```
# --- LOG RETURNS DISTRIBUTION ---
```

```
plt.figure(figsize=(8, 5)) # Set figure size for the histogram
```

```
sns.histplot(stock_data["Log_Return"].dropna(), bins=30, kde=True, color="green") # Plot  
log return distribution in green
```

```
plt.title(f"{symbol} Log Returns Distribution") # Title of the plot
```

```
plt.xlabel("Log Return") # Label for x-axis
```

```
plt.ylabel("Frequency") # Label for y-axis
```

```
plt.grid() # Show grid for better visualization
```

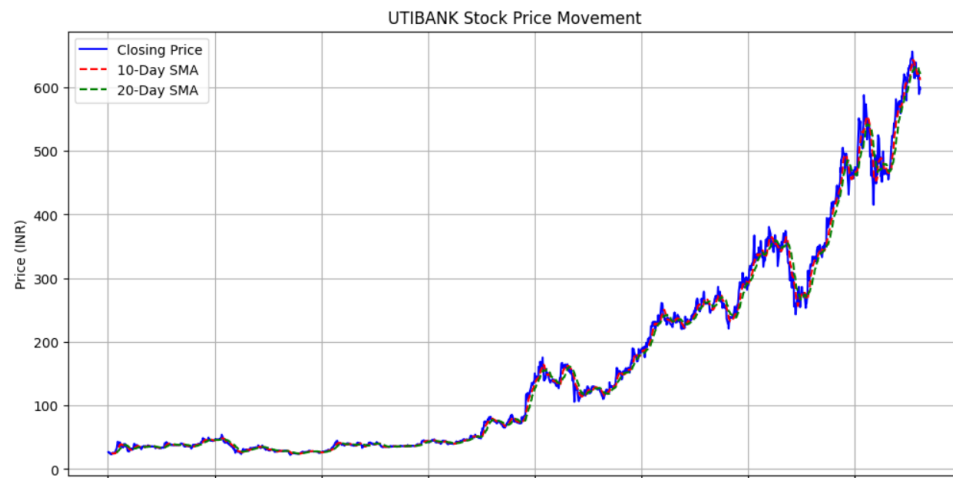
```
plt.show() # Display the plot
```

```

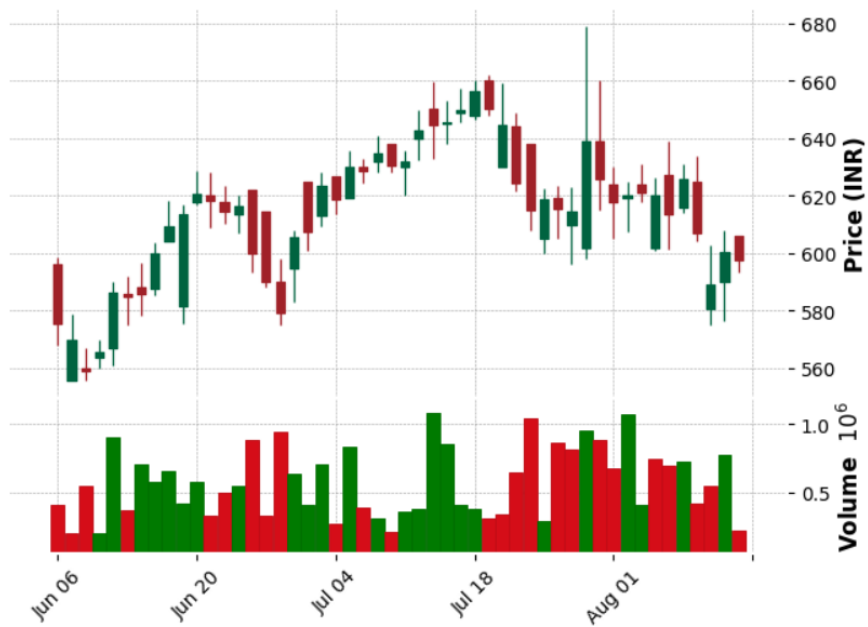
# --- VOLATILITY OVER TIME ---
plt.figure(figsize=(12, 4)) # Set figure size
plt.plot(stock_data["Volatility"], label="Volatility (10-day Rolling)", color="orange") # Plot
volatility in orange
plt.title(f"{symbol} Volatility Over Time") # Title of the plot
plt.xlabel("Date") # Label for x-axis
plt.ylabel("Volatility") # Label for y-axis
plt.legend() # Display legend
plt.grid() # Show grid for better visualization
plt.show() # Display the plot
# --- CANDLESTICK CHART ---
mpf.plot(stock_data[-50:], type="candle", style="charles", # Plot candlestick chart for last
50 days
        title=f"{symbol} Candlestick Chart (Last 50 Days)",
        ylabel="Price (INR)", volume=True) # Volume included in the chart
# Print last 10 rows of key indicators for quick inspection
print(stock_data[["Close", "SMA_10", "SMA_20", "Daily_Return", "Log_Return", "Volatility",
"Signal"]].tail(10))

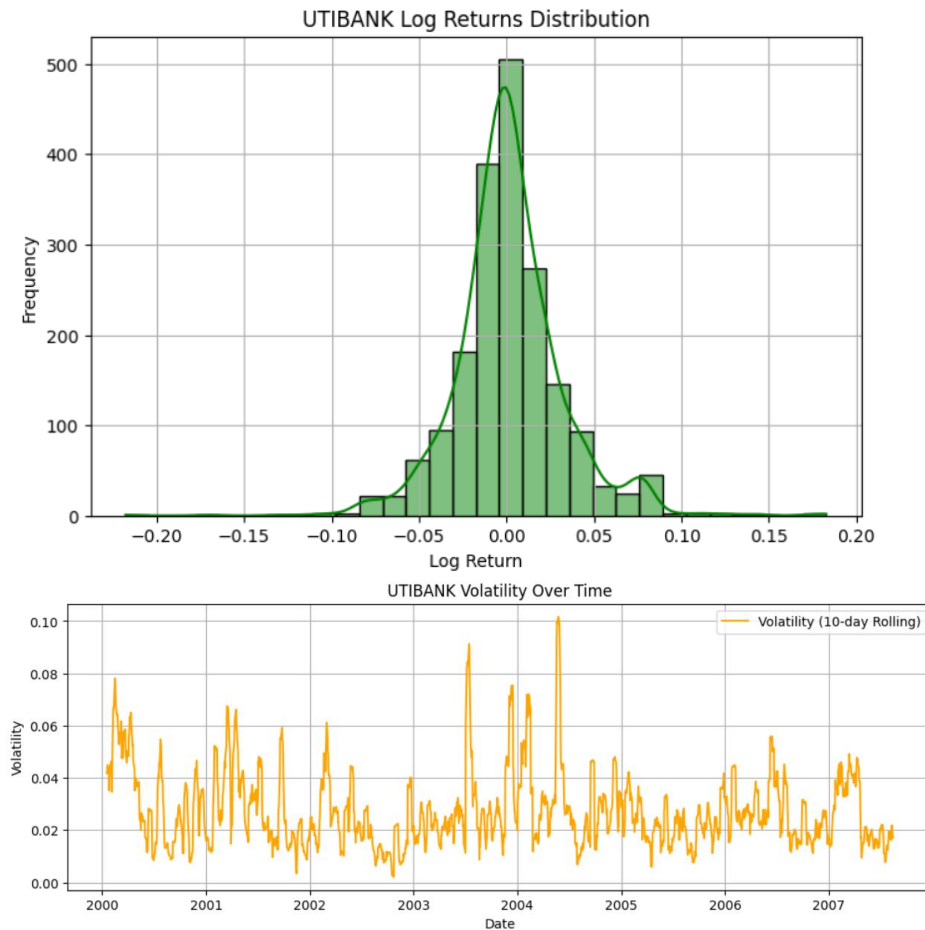
```

#### 4. Output



UTIBANK Candlestick Chart (Last 50 Days)





## 5. Conclusion

This report presents a data-driven approach to analyzing short-term stock price movements. Key insights include:

- Moving averages help in trend identification.
- Daily & log returns measure price fluctuations.
- Volatility analysis highlights market uncertainty.
- Candlestick charts provide a visual representation of price action.

Such analyses assist traders and investors in making informed decisions. Further improvements can include machine learning models for predictive analysis.

## 6. References

Python documentation: <https://docs.python.org/3/>  
Pandas: <https://pandas.pydata.org/>  
Matplotlib: <https://matplotlib.org/>

Seaborn: <https://seaborn.pydata.org/>  
mplfinance: <https://github.com/matplotlib/mplfinance>