

TCS Stock Data Analysis – Project Report

1. Introduction

This project focuses on the analysis of **Tata Consultancy Services (TCS)** stock data using Python. The aim is to understand historical stock price behavior, identify meaningful patterns, and apply basic analytical techniques that are commonly expected from a **Data Analytics Intern**.

The project is designed to reflect a **real-world analytical approach**, keeping the methodology practical, the explanations simple, and the insights business-oriented. Advanced or overly complex techniques have been intentionally avoided to maintain clarity and authenticity.

2. Business Context

Tata Consultancy Services (TCS) is one of India's largest IT services companies and a key constituent of Indian stock market indices. Stock price analysis of such a company provides insights into:

- Long-term growth trends
- Market volatility
- Trading activity and investor behavior

Understanding these aspects is valuable for analysts, investors, and financial institutions.

3. Objective of the Project

The primary objectives of this project are:

- To analyze historical stock price data of TCS
- To perform data cleaning and validation
- To conduct exploratory data analysis (EDA)
- To identify trends using visualizations
- To engineer simple analytical features
- To build a basic predictive model for closing price estimation

This project is **analytical in nature**, not intended for real-time trading or financial advice.

4. Dataset Description

The analysis uses historical stock price data of TCS. The dataset contains daily trading information with the following key columns:

- **Date:** Trading date
- **Open:** Opening price
- **High:** Highest price of the day
- **Low:** Lowest price of the day
- **Close:** Closing price
- **Volume:** Number of shares traded
- **Dividends:** Dividend value (if any)

- **Stock Splits:** Stock split information

The dataset spans multiple years, making it suitable for long-term trend analysis.

5. Tools and Technologies Used

- **Programming Language:** Python
- **Libraries:**
 - Pandas – data manipulation
 - NumPy – numerical operations
 - Matplotlib & Seaborn – visualization
 - Scikit-learn – machine learning
- **Environment:** Jupyter Notebook / VS Code

These tools are widely used in the data analytics industry and suitable for academic as well as internship-level projects.

6. Data Cleaning and Preprocessing

Before analysis, the dataset was prepared carefully:

- Converted the `Date` column to datetime format
- Sorted the data chronologically
- Verified data types for numerical columns
- Checked for missing values (none were found)
- Ensured consistency in financial figures

This step ensured that the dataset was reliable and ready for analysis.

7. Exploratory Data Analysis (EDA)

7.1 Stock Price Trend Analysis

The closing price of TCS stock was plotted over time. The visualization revealed:
- A strong upward trend over the long term
- Periods of volatility during major market events
- Consistent recovery after downturns

This indicates long-term value creation by the company.

7.2 Trading Volume Analysis

Trading volume showed frequent spikes, especially during high-volatility periods. However:

- High volume did not always correspond to sharp price changes
- This suggests institutional trading and long-term investor participation

7.3 Correlation Analysis

A correlation matrix was generated to understand relationships between variables:

Key observations:

- Open, High, Low, and Close prices are highly correlated
- Volume shows weak correlation with price
- Dividends and stock splits have minimal short-term impact on price

This aligns with real-world financial behavior.

8. Feature Engineering

To enhance analysis, additional features were created:

8.1 Moving Averages

- 30-day moving average
- 100-day moving average

These were used to smooth price fluctuations and identify trend direction.

8.2 Daily Percentage Return

Daily percentage change in closing price was calculated to understand volatility and return distribution. The results showed:

- Returns centered around zero
- Presence of extreme values (market shocks)

9. Predictive Modeling

A **Linear Regression model** was implemented to estimate the closing price.

9.1 Model Inputs

Features used:

- Open price
- High price
- Low price
- Trading volume

Target variable:

- Close price

9.2 Model Evaluation

- Mean Squared Error (MSE)
- R-squared score (R^2)

The model performed well due to strong correlation among price variables. However, it is important to note:

- The model is suitable for learning purposes - It does not account for external market factors
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10. Results and Insights

- TCS stock demonstrates strong long-term growth
 - Price-related features are strong predictors of closing price
 - Volume alone is not a reliable indicator of price movement
 - Simple models can explain historical patterns but not future uncertainty
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11. Limitations of the Project

- External factors such as news, global markets, and macroeconomic indicators were not included
- Time-series forecasting models were not explored
- The predictive model is not suitable for real trading decisions

These limitations were consciously accepted to keep the project focused and realistic.

12. Conclusion

This project successfully demonstrates a complete **data analytics workflow**, starting from data understanding to insights and basic prediction. It reflects the practical skills expected from a Data Analytics Intern, including:

- Data cleaning
- Visualization
- Analytical thinking
- Basic modeling

The project maintains a balance between technical depth and business understanding, making it suitable for academic submission and internship evaluation.

13. Future Scope

Possible extensions of this project include: - Time-series forecasting (ARIMA, LSTM) - Inclusion of external financial indicators - Dashboard development using Power BI or Tableau - Risk and volatility analysis

End of Report