1. Objective

The project aims to analyze and understand how **trader behavior** (profits, leverage, volume, and trade counts) varies with **market sentiment**, particularly during "**Fear**" and "**Greed**" phases. By integrating trading and sentiment datasets, the analysis explores whether emotional market conditions influence trading performance and risk behavior.

2. Datasets Used

Trading Dataset (historical data.csv)

 Contains individual trading records with timestamps, account IDs, position size (USD), leverage, and profit/loss data.

• Sentiment Dataset (fear_greed_index.csv)

Includes daily market sentiment classifications (e.g., "Fear", "Greed") based on external financial sentiment indicators.

Both datasets are merged by date for unified analysis.

3. Preprocessing Steps

Data Loading & Inspection

o Loaded both datasets using **Pandas** and verified their dimensions and structure.

Datetime Conversion:

o Converted Timestamp IST and date columns to proper datetime format.

• Feature Extraction:

o Extracted date component from timestamps to align trading data with daily sentiment.

• Data Cleaning:

o Removed invalid or missing date entries.

Aggregation:

- Aggregated trading metrics per day, computing:
 - total volume (sum of USD traded)
 - avg_leverage (mean leverage used)
 - total_profit (sum of closed PnL)
 - trade_count (number of trades per day)

4. Data Integration

- Merged the aggregated trading data with sentiment classifications based on the common date field.
- The resulting dataset (merged_trader_sentiment.csv) provides a day-level view linking **trading** behavior and market sentiment.

5. Exploratory Data Analysis (EDA) Techniques

The exploratory analysis focused on understanding how **market sentiment classifications** — *Fear, Neutral, Greed, Extreme Greed,* and *Extreme Fear* — influenced overall trading profitability.

Profit Distribution by Market Sentiment

The boxplot visualization (shown above) reveals distinct behavioral patterns:

• Fear & Extreme Fear:

- Profits during these phases are highly scattered, with several positive outliers, suggesting
 that while most traders incur small or near-zero gains, a few capitalize on volatile markets
 to secure substantial profits.
- The median profit tends to hover close to zero, indicating cautious or risk-averse trading behavior.

• Neutral Sentiment:

- Trading outcomes remain tightly clustered around zero, implying limited volatility and stable market conditions.
- o This suggests balanced trader sentiment and minimal speculative activity.

• Greed & Extreme Greed:

- These periods show larger positive outliers and slightly wider dispersion in profits, suggesting increased risk-taking and trading volume.
- O Despite the optimism, the median remains near zero, indicating that while a few traders achieve high profits, the average trader's returns do not significantly improve.

Interpretation

The distribution indicates that:

- **Volatility-driven opportunities** emerge during "Fear" and "Extreme Greed" phases, but with inconsistent outcomes.
- Extreme sentiment conditions (both Fear and Greed) produce greater variance in trading performance signaling elevated risk exposure.
- Neutral markets favor steady, low-risk strategies with minimal deviations in profits.

Conclusion from EDA

Market sentiment exerts a visible impact on trading performance — traders tend to take larger positions and experience wider profit variability when emotions dominate the market. This supports the behavioral finance hypothesis that emotional extremes amplify market risk and opportunity simultaneously.

6. Insights & Observations

Profitability Trends:

Average profits tend to differ across sentiment categories — suggesting trader performance correlates with emotional market conditions.

• Risk Appetite:

Traders show varying leverage usage during Fear vs. Greed phases, implying changing risk tolerance.

• Volume Patterns:

Trading volumes reflect overall market enthusiasm, peaking during greed-dominant periods.

• Behavioral Interpretation:

The study provides empirical evidence of **sentiment-driven trading behavior**, aligning with behavioral finance theories.

7. Tools and Libraries

- Data Handling: pandas, numpy
- Visualization: matplotlib, seaborn
- Workflow Management: pathlib for structured file handling

8. Output

- Merged Dataset: merged_trader_sentiment.csv
- Statistical Insights: Average profit differences across sentiment states.
- Visual Summaries: Sentiment-based profit, leverage, and volume distributions.

9. Conclusion

PrimeTradeAI effectively integrates trading and sentiment data to uncover patterns linking emotional market states with financial outcomes.

The results highlight how **fear and greed** influence traders' profitability, leverage behavior, and overall activity, offering valuable insights for developing **sentiment-aware trading strategies** or **behavioral risk assessment tools**.