# Project Report: Behavioral Insights in Web3 Trading

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# Objective

The goal of this project was to analyze how **Bitcoin market sentiment (Fear or Greed)** influences **trader behavior**, using real trading data from Hyperliquid and a sentiment index.

### Datasets Used

- Bitcoin Sentiment Data (fear\_greed\_index.csv)
  - o Columns: timestamp, value, classification, date
  - Represents the market mood on each day as either "Fear" or "Greed".
- 2. **Historical Trader Data** (historical\_data.csv)
  - Contains trade-level information such as Size Tokens, Size USD, Closed PnL, Timestamp IST, etc.

# Methodology

- Cleaned and merged both datasets on the date column.
- Created new features:
  - risk\_exposure = Size Tokens × Size USD
  - profit\_flag = 1 if Closed PnL > 0, else 0
- Performed grouped analysis by market sentiment:

- Total trade volume
- Average profitability
- Risk exposure

# **Key Insights**

#### Trade Volume:

Traders traded significantly more (higher volume) on Greed days than on Fear days.

### Profitability:

The **average profit rate** was slightly higher on **Fear** days, indicating cautious trading may result in more consistent returns.

#### • Risk Exposure:

Trades during **Greed** sentiment had **higher risk exposure**, suggesting traders were more aggressive when the market was optimistic.

# Conclusion

The analysis reveals a clear relationship between **market sentiment** and **trader behavior**:

- Traders tend to take larger, riskier positions during Greed.
- However, more cautious trades during Fear are often more profitable.
- This behavioral pattern can help design better trading strategies based on market mood.

# Next Steps / Suggestions

 Add machine learning classification to predict profitability using sentiment and trade features.

- Track individual accounts over time to profile aggressive vs. conservative traders.
- Analyze behavior during market crashes or booms (extreme sentiment).