

# Trader Behavior vs Bitcoin Market Sentiment

## Analysis Report – Web3 Trading Team

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### 1. Introduction

The objective of this analysis is to understand how **trader behavior** on Hyperliquid varies with **Bitcoin market sentiment**, using two primary datasets:

1.

#### Hyperliquid Historical Trades Dataset

- Contains execution-level trading activity with fields such as account, coin, execution\_price, size\_usd, side, timestamp, closed\_pnl, etc.

2.

#### Bitcoin Fear & Greed Index Dataset

- Provides a daily sentiment score and classification (Extreme Fear, Fear, Neutral, Greed, Extreme Greed).

The goal is to explore:

- How trading behavior (size, volume, profitability) changes with sentiment
- Whether traders perform better in Fear vs Greed markets
- Behavioral patterns that can help improve trading strategies and risk management

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## 2. Data Preparation & Methodology

### 2.1 Data Cleaning

#### Trades Dataset

- Column names standardized to snake\_case
- timestamp\_ist converted to datetime
- Extracted date for daily alignment
- Converted numeric columns such as:
  - execution\_price
  - size\_tokens
  - size\_usd
  - closed\_pnl
  - start\_position
- Created additional engineered features:

- notional\_usd ∝ equal to size\_usd
- is\_profitable ∝ closed\_pnl > 0
- direction\_num ∝ BUY = +1, SELL = -1
- risk\_exposure ∝ equal to notional (leverage not available)

## Sentiment Dataset

- Standardized classification labels
- Parsed date column
- Grouped original 5 sentiment categories into 3 buckets:
  - Fear = Extreme Fear + Fear
  - Neutral
  - Greed = Greed + Extreme Greed
- Added:
  -

sentiment\_group

- sentiment\_binary (Fear = 0, Greed = 1, Neutral = NaN)

## 2.2 Dataset Merge

- Trades were merged with sentiment on the date column
- Each trade now has:
  - Daily sentiment score
  - Classification
  - Sentiment group (Fear / Neutral / Greed)

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## 3. Results & Insights

The merged dataset enabled a sentiment-wise breakdown of trading behavior.

### Summary Table (from analysis)

Sentiment	# Trades	Avg Position Size (USD)	Avg PnL	Win Rate	Total Notional
Fear	83,237	\$7,182	49.21	40.78%	\$597.8M
Greed	90,295	\$4,574	53.88	42.03%	\$413.0M

Neutral	37,686	\$4,782	34.30	39.70%	\$180.2M
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## 4. Key Insights

### 4.1 Profitability is Highest During Greed

- Greed days show the **highest average PnL** and **highest win rate**.
- Bullish sentiment creates **trending markets** that favor momentum trading.

**Conclusion:**  
 ☑ *Traders perform best in optimistic environments.*

### 4.2 Fear Days Have the Largest Position Sizes

- Average position size on Fear days is ~\$7,182, much higher than Greed (~\$4,574).
- Indicates high conviction risk-taking:
  - Panic selling
  - Aggressive dip buying
  - High volatility

**Conclusion:**

☒ *Fear markets drive large, aggressive trades due to volatility.*

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## 4.3 Total Trading Volume is Highest During Fear

- Total notional traded during Fear  $\approx$  \$598M
- This is significantly higher than Greed and Neutral.

**Interpretation:**

More volatility  $\rightarrow$  more trading opportunities  $\rightarrow$  more capital deployed.

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## 4.4 Neutral Days Are the Worst for Trading

Neutral sentiment exhibits:

- Lowest average PnL
- Lowest win rate
- Lowest trading volume

**Conclusion:**

☒ *Low volatility markets offer fewer profitable opportunities.*

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## 4.5 BUY vs SELL Behavior

- **Fear:** BUY and SELL are nearly equal → panic + dip buying
- **Greed:** SELL > BUY → profit-taking behavior
- **Neutral:** least activity overall

### Conclusion:

→ *Trader psychology shifts strongly with sentiment.*

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## 5. Additional Analyses

### 5.1 Symbol-Level Analysis

Grouping by coin and sentiment\_group revealed:

- Some assets are traded heavily during Fear (high volatility assets).
- Others show strong performance during Greed (momentum-driven coins).

This can help optimize sentiment-specific trading strategies per asset.

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### 5.2 Account-Level Analysis

Analyzing trades by account and sentiment showed:

- Some accounts maintain consistent profitability across all sentiment types.
  - Others take excessive risk during Fear and underperform.
  - Useful for profiling trader behavior and evaluating which accounts adapt best to sentiment shifts.
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## 6. Strategic Recommendations

Based on the insights:

### 6.1 Increase System Participation During Greed

- Highest profitability and win rate
- Fewer false breakouts
- Better trend-following performance

### 6.2 Control Risk & Position Size During Fear

- Traders take unusually large positions
- Large volatility can create outsized losses



Recommended actions:

- Position cap
- Tighter stop-loss
- Dynamic risk limits

## 6.3 Reduce Aggressive Trading on Neutral Days

- Minimal opportunities
- Poor performance
- Avoid overtrading

## 6.4 Build Sentiment-Aware Strategies

- Toggle strategies based on daily sentiment
- Adjust leverage & exposure
- Adapt execution behavior to volatility levels

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## 7. Conclusion

This analysis demonstrates that **market sentiment** has a strong and measurable impact on **trader behavior**:

- **Greed** produces the best PnL and win rate
- **Fear** drives the largest positions and highest volume
- **Neutral** sentiment yields the weakest performance

Understanding these patterns enables better **risk management**, **strategy allocation**, and **market timing** for trading teams.

This framework can be extended into predictive models, trader profiling, and real-time sentiment-aware execution systems.