

Trader Behavior vs Bitcoin Market Sentiment

Analysis Report – Web3 Trading Team

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

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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

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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

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:
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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)

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 ≈ \$598M
- This is significantly higher than Greed and Neutral.

Interpretation:

More volatility ☒ more trading opportunities ☒ more capital deployed.

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.*

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.

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

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.