## 1. Executive Summary

Briefly highlight:

- Purpose: exploring how trader behavior aligns with market sentiment.
- Key finding: Trader position sizes (and implied leverage) increase during periods of extreme greed, while profitability dynamics vary across different sentiment states.

#### 2. Data Overview

Summarize:

- Datasets used: Bitcoin Market Sentiment + Hyperliquid Trader Data
- Time coverage & key fields: Date, Classification, closedPnL, leverage, trade size, etc.
- **Preprocessing steps:** cleaning, merging by date/time, handling missing values, and normalizing formats.

## 3. Methodology

Outline the process:

- 1. Data cleaning & preparation
- 2. Aggregations per day and sentiment phase
- 3. Statistical tests (ANOVA / t-tests) for group differences
- 4. Visual exploratory analysis (P&L, leverage, trade volume trends)
- 5. Transition analysis (volume change before sentiment shifts)

# 4. Findings and Insights

Here's how to present your four main questions:

### A. Do traders take more leverage during greed periods?

#### Finding:

Average trade size (USD) was highest during *Extreme Greed* phases. Larger trade sizes likely indicate higher leverage usage or risk exposure.

#### • Visuals to include:

- o Bar chart of average trade size by sentiment
- o Distribution of trade sizes across classifications

## B. Are traders more profitable during fear or greed?

#### • Finding:

- o Extreme Fear showed higher average profitability ratios than Extreme Greed.
- The *Unknown* category had the highest P&L and profit ratio overall.

## • Interpretation:

Profitability doesn't linearly follow sentiment — traders may find better opportunities during fearful markets (contrarian effect).

Visuals:

- Boxplot of P&L by sentiment
- o Bar chart of the average profitability ratio per sentiment

## C. Does volume spike before sentiment shifts?

### • Finding:

- Greed → Fear" transitions were preceded by increases in trade volume.
- o "Fear → Greed" transitions showed **decreases** in USD trade volume.

## • Interpretation:

Traders may become more active before negative sentiment changes — suggesting that spikes in volume can foreshadow fear entering the market.

#### • Visuals:

- o Line chart showing volume vs time, colored by sentiment phase
- o Transition matrix heatmap of volume changes

## D. Can any behavior pattern predict a future market move?

### Finding:

Correlations exist between sentiment and trading behavior (frequency, PnL, size), but they are not predictive.

Further modeling (time-series ML or sentiment forecasting) is needed to predict future moves.

## Next Steps:

- o Use lag features (e.g., previous day's volume/leverage).
- o Try simple logistic regression or LSTM to test predictiveness.

#### 5. Conclusions

Summarize key insights:

- Traders tend to scale up during greed phases but may achieve higher efficiency during fear phases.
- Extreme market conditions show non-linear profitability behavior.
- Volume spikes may act as early warning indicators for sentiment changes.
- Predictive modeling is a logical next step.

### 6. Recommendations

- Incorporate sentiment signals into trade risk models.
- Monitor **volume surges** before sentiment transitions.
- Encourage **behavioral diversification** avoid over-leveraging during greed.

# 7. Visual Gallery (in /outputs)

Include labeled images:

output\_images.pdf