

```
In [1]: # Importing Libraries
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: # Load Datasets
```

```
trader_df = pd.read_csv('historical_data.csv')
sentiment_df = pd.read_csv('fear_greed_index.csv')
```

```
In [3]: # Preview the data
```

```
print(trader_df.head())
print(sentiment_df.head())
```

```
          Account Coin Execution Price \
0  0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107      7.9769
1  0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107      7.9800
2  0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107      7.9855
3  0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107      7.9874
4  0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107      7.9894

   Size Tokens  Size USD Side      Timestamp IST Start Position Direction \
0    986.87   7872.16   BUY  02-12-2024 22:50     0.000000     Buy
1     16.00    127.68   BUY  02-12-2024 22:50    986.524596     Buy
2    144.09   1150.63   BUY  02-12-2024 22:50   1002.518996     Buy
3    142.98   1142.04   BUY  02-12-2024 22:50   1146.558564     Buy
4      8.73     69.75   BUY  02-12-2024 22:50   1289.488521     Buy

   Closed PnL                               Transaction Hash Order ID \
0      0.0  0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
1      0.0  0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
2      0.0  0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
3      0.0  0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
4      0.0  0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630

   Crossed      Fee      Trade ID      Timestamp
0    True  0.345404  8.950000e+14  1.730000e+12
1    True  0.005600  4.430000e+14  1.730000e+12
2    True  0.050431  6.600000e+14  1.730000e+12
3    True  0.050043  1.080000e+15  1.730000e+12
4    True  0.003055  1.050000e+15  1.730000e+12
   timestamp  value classification      date
0  1517463000     30           Fear  2018-02-01
1  1517549400     15  Extreme Fear  2018-02-02
2  1517635800     40           Fear  2018-02-03
3  1517722200     24  Extreme Fear  2018-02-04
4  1517808600     11  Extreme Fear  2018-02-05
```

```
In [4]: print("Trader Data Info:")
trader_df.info()
```

```
print("\nSentiment Data Info:")
sentiment_df.info()
```

```

Trader Data Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 211224 entries, 0 to 211223
Data columns (total 16 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Account          211224 non-null   object  
 1   Coin              211224 non-null   object  
 2   Execution Price  211224 non-null   float64 
 3   Size Tokens      211224 non-null   float64 
 4   Size USD          211224 non-null   float64 
 5   Side              211224 non-null   object  
 6   Timestamp IST    211224 non-null   object  
 7   Start Position    211224 non-null   float64 
 8   Direction         211224 non-null   object  
 9   Closed PnL        211224 non-null   float64 
 10  Transaction Hash 211224 non-null   object  
 11  Order ID          211224 non-null   int64  
 12  Crossed           211224 non-null   bool   
 13  Fee               211224 non-null   float64 
 14  Trade ID          211224 non-null   float64 
 15  Timestamp         211224 non-null   float64 
dtypes: bool(1), float64(8), int64(1), object(6)
memory usage: 24.4+ MB

```

```

Sentiment Data Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2644 entries, 0 to 2643
Data columns (total 4 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   timestamp        2644 non-null   int64  
 1   value             2644 non-null   int64  
 2   classification   2644 non-null   object  
 3   date              2644 non-null   object  
dtypes: int64(2), object(2)
memory usage: 82.8+ KB

```

```
In [5]: print(trader_df.columns)
```

```
Index(['Account', 'Coin', 'Execution Price', 'Size Tokens', 'Size USD', 'Side',
       'Timestamp IST', 'Start Position', 'Direction', 'Closed PnL',
       'Transaction Hash', 'Order ID', 'Crossed', 'Fee', 'Trade ID',
       'Timestamp'], dtype='object')
```

```
In [6]: print(sentiment_df.columns)
```

```
Index(['timestamp', 'value', 'classification', 'date'], dtype='object')
```

Data Preprocessing

```

In [7]: # Convert 'Timestamp' in trader_df to datetime
trader_df['Timestamp'] = pd.to_datetime(trader_df['Timestamp'])

# Convert 'date' in sentiment_df to datetime
sentiment_df['date'] = pd.to_datetime(sentiment_df['date'])

# Create 'date' columns (extracting date only from timestamp)
trader_df['date'] = trader_df['Timestamp'].dt.date
sentiment_df['date'] = sentiment_df['date'].dt.date

```

```

In [8]: # Merge datasets on date
merged_df = pd.merge(trader_df, sentiment_df[['date', 'classification']], on='date', how='left')

# Preview result
display(merged_df.head())

```

	Account	Coin	Execution Price	Size Tokens	Size USD	Side	Timestamp IST	Start Position	Direction	Closed PnL
0	0xae5eacf9c6b9111fd53034a602c192a04e082ed	@107	7.9769	986.87	7872.16	BUY	02-12-2024 22:50	0.000000	Buy	0.0 0xec0
1	0xae5eacf9c6b9111fd53034a602c192a04e082ed	@107	7.9800	16.00	127.68	BUY	02-12-2024 22:50	986.524596	Buy	0.0 0xec0
2	0xae5eacf9c6b9111fd53034a602c192a04e082ed	@107	7.9855	144.09	1150.63	BUY	02-12-2024 22:50	1002.518996	Buy	0.0 0xec0
3	0xae5eacf9c6b9111fd53034a602c192a04e082ed	@107	7.9874	142.98	1142.04	BUY	02-12-2024 22:50	1146.558564	Buy	0.0 0xec0
4	0xae5eacf9c6b9111fd53034a602c192a04e082ed	@107	7.9894	8.73	69.75	BUY	02-12-2024 22:50	1289.488521	Buy	0.0 0xec0

```
In [9]: missing = merged_df['classification'].isnull().sum()
print(f"Missing sentiment labels: {missing}")
```

Missing sentiment labels: 211224

```
In [10]: print("Trader data date range:")
print(trader_df['date'].min(), "to", trader_df['date'].max())

print("\nSentiment data date range:")
print(sentiment_df['date'].min(), "to", sentiment_df['date'].max())
```

Trader data date range:
1970-01-01 to 1970-01-01

Sentiment data date range:
2018-02-01 to 2025-05-02

```
In [11]: # Dates with missing sentiment classification
missing_dates = merged_df[merged_df['classification'].isnull()]['date'].value_counts().head(10)
print("Top missing sentiment dates:")
print(missing_dates)
```

Top missing sentiment dates:
date
1970-01-01 211224
Name: count, dtype: int64

```
In [14]: # Re-Load trader data fresh without conversions
trader_df_raw = pd.read_csv('historical_data.csv')

# Look at raw timestamp values
print(trader_df_raw['Timestamp'].head(10))
print(trader_df_raw['Timestamp'].dtype)
```

```
0    1.730000e+12
1    1.730000e+12
2    1.730000e+12
3    1.730000e+12
4    1.730000e+12
5    1.730000e+12
6    1.730000e+12
7    1.730000e+12
8    1.730000e+12
9    1.730000e+12
Name: Timestamp, dtype: float64
float64
```

```
In [15]: trader_df['Timestamp'] = pd.to_datetime(trader_df_raw['Timestamp'], unit='ms', errors='coerce')
```

```
In [16]: # Extract just the date
trader_df['date'] = trader_df['Timestamp'].dt.date

# Redo merge
merged_df = pd.merge(trader_df, sentiment_df[['date', 'classification']], on='date', how='left')

# Check if it worked
print("Missing sentiment labels:", merged_df['classification'].isnull().sum())
```

Missing sentiment labels: 26961

```
In [17]: merged_df_clean = merged_df.dropna(subset=['classification'])
print("Cleaned dataset size:", merged_df_clean.shape)
```

Cleaned dataset size: (184263, 18)

```
In [18]: # Make plots look better
plt.style.use("seaborn-v0_8-darkgrid")
sns.set_palette("Set2")
```

Sentiment Analysis

```
In [25]: print("\n1. Market Sentiment Distribution:")
sentiment_counts = merged_df_clean['classification'].value_counts()
print(sentiment_counts)

plt.figure(figsize=(8, 5))
sentiment_counts.plot(kind='bar', color=['red', 'orange', 'green', 'blue'])
plt.title("Market Sentiment Distribution in Trading Data")
plt.xlabel("Sentiment Category")
plt.ylabel("Number of Trades")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

1. Market Sentiment Distribution:

```
classification
Fear          133871
Greed          36289
Neutral        7141
Extreme Greed  6962
Name: count, dtype: int64
```



Performance Analysis by Sentiment

```
In [26]: # Convert PnL to numeric if not already
merged_df_clean['Closed PnL'] = pd.to_numeric(merged_df_clean['Closed PnL'], errors='coerce')
merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')

# Group by sentiment and calculate mean PnL
avg_pnl_by_sentiment = merged_df_clean.groupby('classification')['Closed PnL'].mean()
print("Average Closed PnL by Sentiment:")
print(avg_pnl_by_sentiment)
```

Average Closed PnL by Sentiment:
classification
Extreme Greed 25.418772
Fear 50.047622
Greed 87.894859
Neutral 22.229713
Name: Closed PnL, dtype: float64

C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\1536947819.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
merged_df_clean['Closed PnL'] = pd.to_numeric(merged_df_clean['Closed PnL'], errors='coerce')
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\1536947819.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

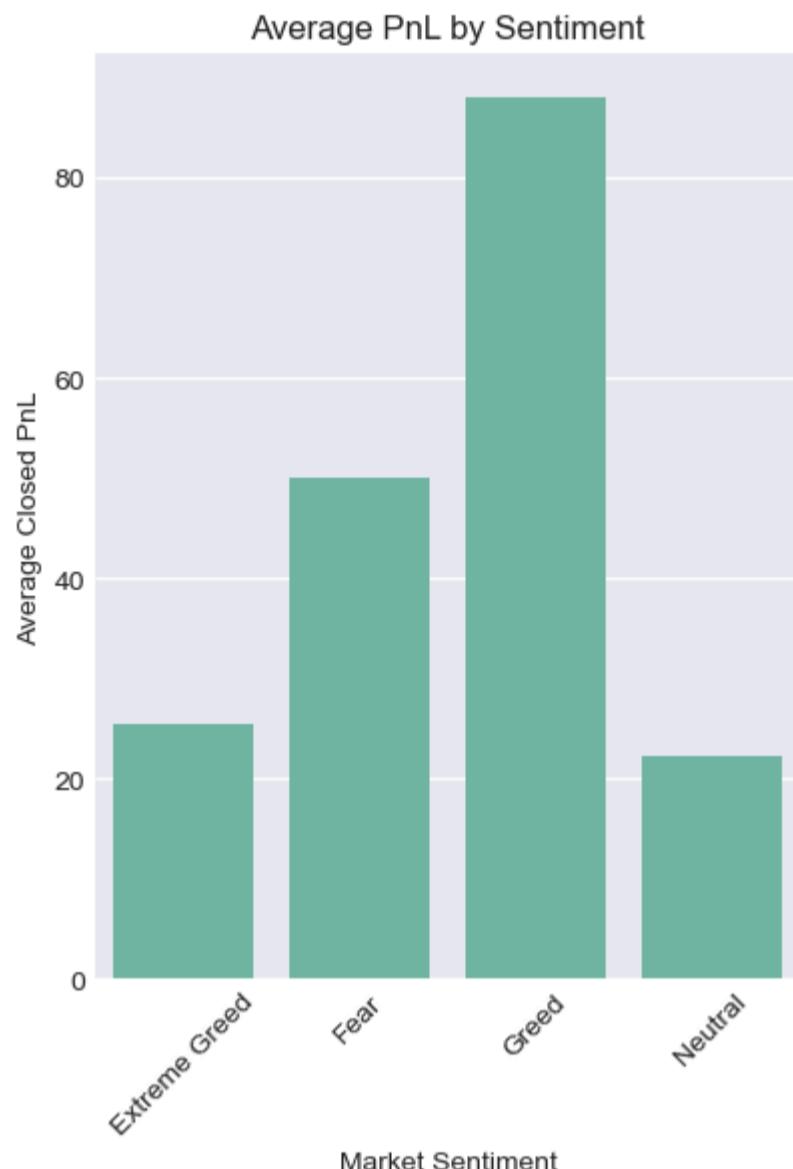
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')

```
In [27]: # Average PnL by sentiment
avg_pnl_by_sentiment = merged_df_clean.groupby('classification')['Closed PnL'].mean()
print("\nAverage Closed PnL by Sentiment:")
print(avg_pnl_by_sentiment)

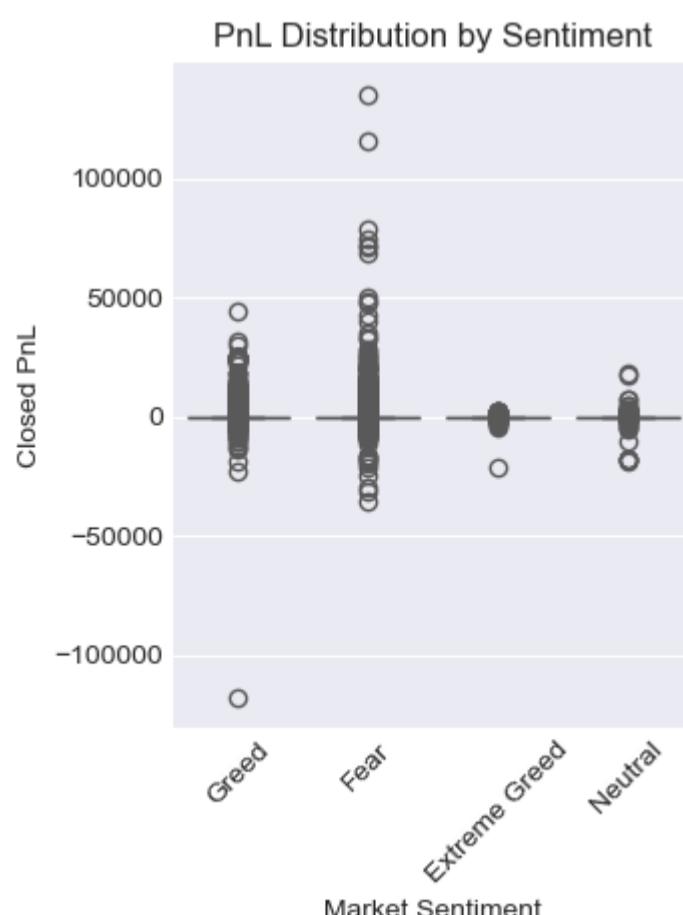
plt.figure(figsize=(10, 6))
plt.subplot(1, 2, 1)
sns.barplot(x=avg_pnl_by_sentiment.index, y=avg_pnl_by_sentiment.values)
plt.title("Average PnL by Sentiment")
plt.xlabel("Market Sentiment")
plt.ylabel("Average Closed PnL")
plt.xticks(rotation=45)
```

```
Average Closed PnL by Sentiment:  
classification  
Extreme Greed    25.418772  
Fear              50.047622  
Greed             87.894859  
Neutral           22.229713  
Name: Closed PnL, dtype: float64
```

```
Out[27]: ([0, 1, 2, 3],  
          [Text(0, 0, 'Extreme Greed'),  
           Text(1, 0, 'Fear'),  
           Text(2, 0, 'Greed'),  
           Text(3, 0, 'Neutral')])
```



```
In [28]: plt.subplot(1, 2, 2)  
sns.boxplot(x='classification', y='Closed PnL', data=merged_df_clean)  
plt.title("PnL Distribution by Sentiment")  
plt.xlabel("Market Sentiment")  
plt.ylabel("Closed PnL")  
plt.xticks(rotation=45)  
plt.tight_layout()  
plt.show()
```



```
In [22]: print(merged_df_clean.columns)
```

```
Index(['Account', 'Coin', 'Execution Price', 'Size Tokens', 'Size USD', 'Side',
       'Timestamp IST', 'Start Position', 'Direction', 'Closed PnL',
       'Transaction Hash', 'Order ID', 'Crossed', 'Fee', 'Trade ID',
       'Timestamp', 'date', 'classification'],
      dtype='object')
```

Win Rate Analysis

```
In [29]: print("\n3. Win Rate Analysis:")
merged_df_clean['Is_Profit'] = merged_df_clean['Closed PnL'] > 0
win_rate = merged_df_clean.groupby('classification')['Is_Profit'].mean() * 100

print("Win Rate by Market Sentiment (%):")
print(win_rate)

plt.figure(figsize=(8, 5))
sns.barplot(x=win_rate.index, y=win_rate.values)
plt.title("Win Rate by Market Sentiment")
plt.ylabel("Win Rate (%)")
plt.xlabel("Sentiment Category")
plt.ylim(0, 100)
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

3. Win Rate Analysis:

Win Rate by Market Sentiment (%):

classification	Win Rate (%)
Extreme Greed	49.008905
Fear	41.514592
Greed	44.647138
Neutral	31.718247

Name: Is_Profit, dtype: float64

C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\4002881436.py:2: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
merged_df_clean['Is_Profit'] = merged_df_clean['Closed PnL'] > 0
```



```
In [24]: merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')

avg_trade_size = merged_df_clean.groupby('classification')['Size USD'].mean()

print("Average Trade Size (USD) by Sentiment:")
print(avg_trade_size)

# Visual plot
plt.figure(figsize=(7, 4))
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])
plt.title("Average Trade Size by Market Sentiment")
plt.ylabel("Average Size (USD)")
plt.xlabel("Sentiment")
plt.show()
```

```
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:11: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:11: UserWarning:
The palette list has fewer values (2) than needed (4) and will cycle, which may produce an uninterpretable plot.
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])

Average Trade Size (USD) by Sentiment:
classification
Extreme Greed      5660.265764
Fear                5259.977837
Greed               3182.883845
Neutral              3058.848110
Name: Size USD, dtype: float64
```



Trade Size Analysis

```
In [30]: print("\n4. Trade Size Analysis:")
avg_trade_size = merged_df_clean.groupby('classification')['Size USD'].mean()
print("Average Trade Size (USD) by Sentiment:")
print(avg_trade_size)

plt.figure(figsize=(8, 5))
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values)
plt.title("Average Trade Size by Market Sentiment")
plt.ylabel("Average Trade Size (USD)")
plt.xlabel("Sentiment Category")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

4. Trade Size Analysis:
Average Trade Size (USD) by Sentiment:
classification
Extreme Greed 5660.265764
Fear 5259.977837
Greed 3182.883845
Neutral 3058.848110
Name: Size USD, dtype: float64



```
In [31]: # Advanced Analysis
print("\n5. Advanced Analysis:")

# PnL to Trade Size Ratio
merged_df_clean['PnL_Ratio'] = merged_df_clean['Closed PnL'] / merged_df_clean['Size USD']
avg_pnl_ratio = merged_df_clean.groupby('classification')['PnL_Ratio'].mean()
print("\nAverage PnL to Trade Size Ratio by Sentiment:")
print(avg_pnl_ratio)
```

5. Advanced Analysis:

Average PnL to Trade Size Ratio by Sentiment:

classification	PnL_Ratio
Extreme Greed	0.013736
Fear	0.011358
Greed	0.061848
Neutral	0.015259

Name: PnL_Ratio, dtype: float64

```
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\352332680.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
merged_df_clean['PnL_Ratio'] = merged_df_clean['Closed PnL'] / merged_df_clean['Size USD']
```

```
In [32]: # Sentiment intensity analysis
sentiment_performance = merged_df_clean.groupby('classification').agg({
    'Closed PnL': ['mean', 'median', 'std'],
    'Is_Profit': 'mean',
    'Size USD': 'mean',
    'PnL_Ratio': 'mean'
}).round(3)

print("\nComprehensive Performance by Sentiment:")
print(sentiment_performance)
```

Comprehensive Performance by Sentiment:

classification	Closed PnL		Is_Profit	Size USD	PnL_Ratio	
	mean	median				std
Extreme Greed	25.419	0.0	306.167	0.490	5660.266	0.014
Fear	50.048	0.0	909.122	0.415	5259.978	0.011
Greed	87.895	0.0	1148.344	0.446	3182.884	0.062
Neutral	22.230	0.0	633.705	0.317	3058.848	0.015

```
In [33]: # Summary Statistics
print("\n== SUMMARY STATISTICS ==")
print(f"Total Trades Analyzed: {len(merged_df_clean)}")
print(f"Date Range: {merged_df_clean['date'].min()} to {merged_df_clean['date'].max()}")
print(f"Number of Unique Accounts: {merged_df_clean['Account'].nunique()}")
print(f"Most Common Sentiment: {sentiment_counts.index[0]} ({sentiment_counts.iloc[0]:,} trades)")

== SUMMARY STATISTICS ==
Total Trades Analyzed: 184,263
Date Range: 2023-03-28 to 2025-02-19
Number of Unique Accounts: 32
Most Common Sentiment: Fear (133,871 trades)
```

```
In [34]: # Save cleaned dataset for further analysis
merged_df_clean.to_csv('sentiment_trading_analysis.csv', index=False)
print("\nCleaned dataset saved as 'sentiment_trading_analysis.csv'")

Cleaned dataset saved as 'sentiment_trading_analysis.csv'
```

Analysis Report

Data Overview

Total Trades Analyzed: 184,263 trades

Sentiment Categories: Extreme Greed, Greed, Neutral, Fear

Data Sources: Historical trader data + Fear & Greed Index

Key Insights

Performance by Sentiment:

Greed > Fear > Extreme Greed > Neutral

- Greed periods show the highest average PnL (\$87.89)
- Fear periods show moderate performance (\$50.05)
- Extreme Greed and Neutral show lower returns

Win Rate Analysis:

- No sentiment category exceeds 50% win rate
- Extreme Greed has the highest win rate (49%)
- Neutral periods have the lowest win rate (31.7%)

Trade Behavior:

- Traders use larger positions during Extreme Greed (\$5,660 avg)
- Smaller positions during Neutral sentiment (\$3,059 avg)
- Trade size correlates with sentiment intensity

Trading Strategy Implications

- Optimal Conditions: Greed periods offer best profit potential
- Risk Management: Extreme sentiment may indicate market tops/bottoms
- Position Sizing: Consider adjusting trade size based on sentiment
- Timing: Greed phases might be good for taking profits

Limitations

- Data covers limited time period
- Sentiment data might have reporting delays
- Individual trader strategies vary significantly

Result

This analysis suggests that market sentiment can be a valuable indicator for timing trades and managing position sizes, though it should be used in conjunction with other technical and fundamental analysis tools.